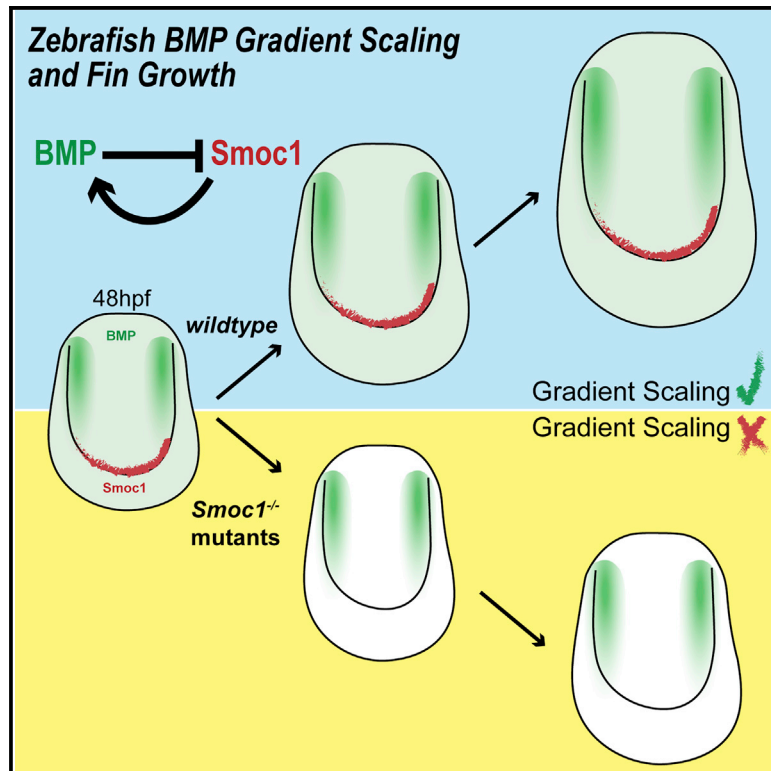# BMP Signaling Gradient Scaling in the Zebrafish Pectoral Fin

Graphical Abstract

Authors

Rita Mateus, Laurent Holtzer,
Carole Seum, Zena Hadjivasiliou,
Marine Dubois, Frank Jülicher,
Marcos Gonzalez-Gaitan

Correspondence

marcos.gonzalez@unige.ch

In Brief

The control of organ growth is essential for embryonic development. Mateus et al. show that during zebrafish pectoral fin development, two BMP morphogen signaling gradients form and expand proportionally to fin size, regulating growth. A feedback loop between Smoc1 (a conserved secreted factor) and BMP signaling mediates BMP gradient scaling.

Highlights

- Zebrafish pectoral fins form two BMP signaling gradients that scale with fin size

- The gradients show signatures of growth control by time derivatives of signaling

- Smoc1 mutants fail to scale the BMP signaling gradient and have fin growth defects

- A regulatory feedback loop Smoc1-BMP mediates BMP gradient scaling

CellPress

BMP Signaling Gradient Scaling in the Zebrafish Pectoral Fin

Rita Mateus,[1,3] Laurent Holtzer,[1,3] Carole Seum,[1] Zena Hadjivasiliou,[1] Marine Dubois,[1] Frank Jülicher,[2] and Marcos Gonzalez-Gaitan[1,4,*]
[1]Department of Biochemistry, Faculty of Sciences, University of Geneva, Switzerland
[2]Max Planck Institute for the Physics of Complex Systems, Dresden, Germany
[3]These authors contributed equally
[4]Lead Contact
*Correspondence: marcos.gonzalez@unige.ch
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SUMMARY

Secreted growth factors can act as morphogens that form spatial concentration gradients in developing organs, thereby controlling growth and patterning. For some morphogens, adaptation of the gradients to tissue size allows morphological patterns to remain proportioned as the organs grow. In the zebrafish pectoral fin, we found that BMP signaling forms a two-dimensional gradient. The length of the gradient scales with tissue length and its amplitude increases with fin size according to a power-law. Gradient scaling and amplitude power-laws are signatures of growth control by time derivatives of morphogenetic signaling: cell division correlates with the fold change over time of the cellular signaling levels. We show that Smoc1 regulates BMP gradient scaling and growth in the fin. Smoc1 scales the gradient by means of a feedback loop: Smoc1 is a BMP agonist and BMP signaling represses Smoc1 expression. Our work uncovers a layer of morphogen regulation during vertebrate appendage development.

INTRODUCTION

During embryogenesis, morphogen gradients control patterning as well as growth of tissues and organs (Crick, 1970; Turing, 1952; Wolpert, 1969). Some morphogens are growth factors expressed from a restricted group of cells in developing tissues, from where they are secreted and form concentration gradients. In some cases, they endow cells with positional information by activating different target genes at different morphogen concentration thresholds. The shape of the gradient patterns the tissue.

One remarkable feature of some morphogen gradients is their ability to adapt spatially to tissues of different sizes: the gradient shape is invariant and scales to keep the morphological patterns proportional, independent of organ size. Not all morphogen gradients scale (see for example the Hedgehog gradient in the wing [Wartlick et al., 2011]) and, as already discussed by Wolpert, how gradients could possibly scale is not a trivial issue (Wolpert, 1969). But morphogenetic gradient scaling has indeed been

shown to regulate different aspects of scale invariant patterning during animal development in planarians, insects, and vertebrates (Almuedo-Castillo et al., 2018; Ben-Zvi et al., 2008; Huang and Umulis, 2019; Stückemann et al., 2017; Teleman and Cohen, 2000; Wartlick et al., 2011). Importantly, it has been proposed that gradient scaling can also control tissue growth. In insects, cells sense the relative fold increase in bone morphogenetic protein (BMP) signaling over time, so that cells divide when a signaling threshold is met (Wartlick et al., 2011, 2014). In this context, gradient scaling can mediate the increase in the cellular signaling levels over time, which is necessary for growth.

The mechanisms underlying gradient scaling are not yet understood. Different studies have proposed that gradient scaling can be achieved by morphogen advection-dilution or by more complex processes such as two opposing gradients, shuttling, the interplay between a source and a sink, or by means of an expander-repression mechanism (Averbukh et al., 2014; Ben-Zvi and Barkai, 2010; McHale et al., 2006; Shilo et al., 2013; Zinski et al., 2017). In particular, the expansion-repression mechanism involves an expander molecule that relays information about the size of the tissue into the morphogen gradient, allowing for its scaling. In this model, the expander transcription is repressed by the morphogen and is thereby restricted to the edge of the gradient (Ben-Zvi and Barkai, 2010). From there, the expander diffuses throughout the target tissue and promotes the expansion of the gradient by regulating morphogen degradation and/or diffusion. The fact that the expander is expressed at the edge of the gradient and the tissue, encodes the necessary information to compare the length of the gradient with the size of the tissue: expander expression increases if the tissue is bigger than the gradient. A candidate molecule for an expander has been proposed in insects, Pentagone (Vuilleumier et al., 2010). Pentagone is indeed expressed at the edge of a BMP gradient, controls degradation of the ligand, and is essential for the scaling of the gradient (Ben-Zvi et al., 2011; Hamaratoglu et al., 2011; Wartlick et al., 2014).

In vertebrates, during embryonic dorso-ventral axis specification, TGFβ family ligands are also distributed as gradients of concentration. In particular, in *Xenopus* and zebrafish, both BMP and Nodal form gradients which mediate patterning and scale, adapting to embryo size (Almuedo-Castillo et al., 2018; Ben-Zvi et al., 2008; Huang and Umulis, 2019; Reversade and De Robertis, 2005). Instead of looking at embryos of different sizes, here we study how a gradient expands and scales in an organ as

it grows. In particular, we investigate BMP signaling in the zebrafish pectoral fin during its proliferation phase in the embryo (Ning et al., 2013; Yano et al., 2012). The pectoral fin is the homolog of the tetrapod forelimbs, being initially comprised by a mesenchymal bud and surrounding ectodermal cells. Later, while the fin extends its proximal-distal axis, it defines two regions: the endoskeletal disc proximally and the apical fold distally (Grandel and Schulte-Merker, 1998; Mercader, 2007). Despite extensive knowledge about the contribution of signaling molecules necessary to establish these patterns, the mechanisms controlling fin growth and its coordination with patterning are not understood (Mercader, 2007; Yano et al., 2012).

Here we propose that BMP signaling gradients regulate pectoral fin growth by means of gradient scaling, mediated by an expansion-repression mechanism. In the fin, we found that two BMP signaling gradients are established and scale in two dimensions with the fin's size. Importantly, we show that gradient scaling is coupled to the relative increase in BMP signaling over time, which could underlie the regulation of cell division and organ size. Furthermore, we identify a vertebrate expander molecule, Smoc1, which participates in a regulatory feedback loop with the BMP gradients. BMP gradient scaling in the pectoral fin is fine-tuned through an expansion-repression mechanism, allowing for gradient scaling during fin growth.

RESULTS

Two BMP Signaling Gradients during Pectoral Fin Growth

In zebrafish, previous studies have shown that BMP signaling regulates embryonic fin growth (Ning et al., 2013; Figures S1A–S1I). To study the dynamics of BMP signaling during this process, we imaged transgenic fish carrying reporter constructs based on a BMP responsive element (BRE; Figure 1; Video S1). These BRE reporters have been previously used as live sensors of BMP signaling in zebrafish, given that GFP is expressed under the control of Smad1/5/8 enhancers from the mouse Id1 promoter (Collery and Link, 2011; Korchynskyi and ten Dijke, 2002; Laux et al., 2011).

During pectoral fin development, we found that BMP signaling forms two spatial concentration gradients that decay along the proximo-distal axis, in the anterior and posterior sides of the endoskeletal disc (Figures 1A–1E). This signaling pattern was confirmed by anti-Phospho-Smad1/5/9 immunostainings (Figures S1J–S1L; Collery and Link, 2011). Consistent with the formation of proximo-distal signaling gradients, the BMP2a ligand has been shown to be expressed at the proximal base of the fin (Martínez-Barberá et al., 1997; Neumann et al., 1999) and the BmpR1b receptor is present throughout the endoskeletal disc (Figures S1M–S1O). In the center of the endoskeletal disc, lower levels of signaling correlate with the expression of Noggin, a BMP inhibitor that binds the ligand, thereby competing with the receptor (Fürthauer et al., 1999; Zimmerman et al., 1996). Therefore, in addition to the proximo-distal decays, the signaling gradients also decay toward the center of the endoskeletal disc (Figures S1P–S1S).

The fin primordium is a 3D object with dorso-ventral (DV), antero-posterior (AP), and proximo-distal (PD) axes. Since the fin

endoskeletal disc is thin along the DV axis (Figure S2A), we projected the BRE signal to consider only the PD/AP plane. In this surface, the BRE signal $C(s)$ fit exponential curves $C(s) = C_0 e^{-s/\lambda}$, with s, the position along the gradient; λ, its length scale; and C_0, its amplitude, i.e., the maximal concentration at the proximal side of the fin. An exponential profile captures the shape of the gradient, whether we considered the distance s along the gradient (ROI in Figures 1B, 1D, 1F, and 1G) or the position (x, y) along the PD and AP axes (Figures S2D–S2J; see STAR Methods and Supplemental Methods for Theoretical Framework).

The anterior and posterior gradients have approximately the same length scale of about 5 cell diameters (around 28 μm) at 60 h post-fertilization (hpf; Figures 1F and 1G). The amplitude of the anterior gradient is smaller than that of the posterior (Figures 1E, S2B, and S2C). This is consistent with previous reports, showing that BMP2a expression at the base of the fin is stronger in the posterior side (Martínez-Barberá et al., 1997; Neumann et al., 1999).

The two gradients can be observed from 48 to 78 hpf, a larval period in which the pectoral fin grows actively, doubling its proximo-distal length (Figures 2A–2F and 2M). Before 48 hpf, the gradients are too faint to quantify reliably and after 78 hpf, the profile does not fit an exponential curve, due to high levels of expression in single mesenchymal cells (Figures S1T and S1U).

Fin Growth Is Homogeneous and Anisotropic

To study the BMP gradients as the fin grows, we first characterized fin growth by scanning electron microscopy (Figures 2A–2F and 2M) and cell proliferation, by imaging Fucci live sensors (S/G2/M marker) (Sugiyama et al., 2009) as well as performing phospho-histone H3 immunostainings (mitosis marker) (Hans and Dimitrov, 2001) (Figures 2G–2L and 2N). We observed that, from 48 to 95 hpf, the proliferation rate decays exponentially, with growth stopping around 78 hpf (Figure 2N).

The even distribution of cells in S/G2/M throughout the fin primordium (Figures 2J–2L) indicates that proliferation is approximately homogeneous until 72 hpf; growth ceases around this time (Figure 2N), when some heterogeneity can be observed (Figure 2L), as previously reported (Freitas et al., 2012; Grandel and Schulte-Merker, 1998). Dorso-ventral fin growth is negligible (Figure S2A), and growth is anisotropic along the two other axes, with a larger PD growth rate: $\varepsilon = (g_{AP}/g_{PD}) = 0.65 \pm 0.024$ (Figures 2M and 2O). While fin proliferation is approximately uniform, BMP signaling is graded, indicating that the growth rate does not correlate with the levels of signaling. Instead, previous reports have shown that the dynamics of BMP signaling can control cell proliferation (Wartlick et al., 2011, 2014). For this reason, we studied next the dynamics of signaling as the fin develops.

The BMP Gradients Scale with the Size of the Fin

As the fin grows, the decay lengths of both anterior and posterior BRE gradients expand. The anterior and posterior decay lengths are approximately the same and both are proportional to tissue length L (half-length along the ROI midline): $\phi_A = (\lambda_A/L) = 0.31 \pm 0.04$ and $\phi_P = (\lambda_P/L) = 0.35 \pm 0.05$ (Figures 3A–3D and S2D–S2J). We observed similar gradient scaling with fin size with anti-Phospho-Smad1/5/9 immunostainings (Figures S4I,
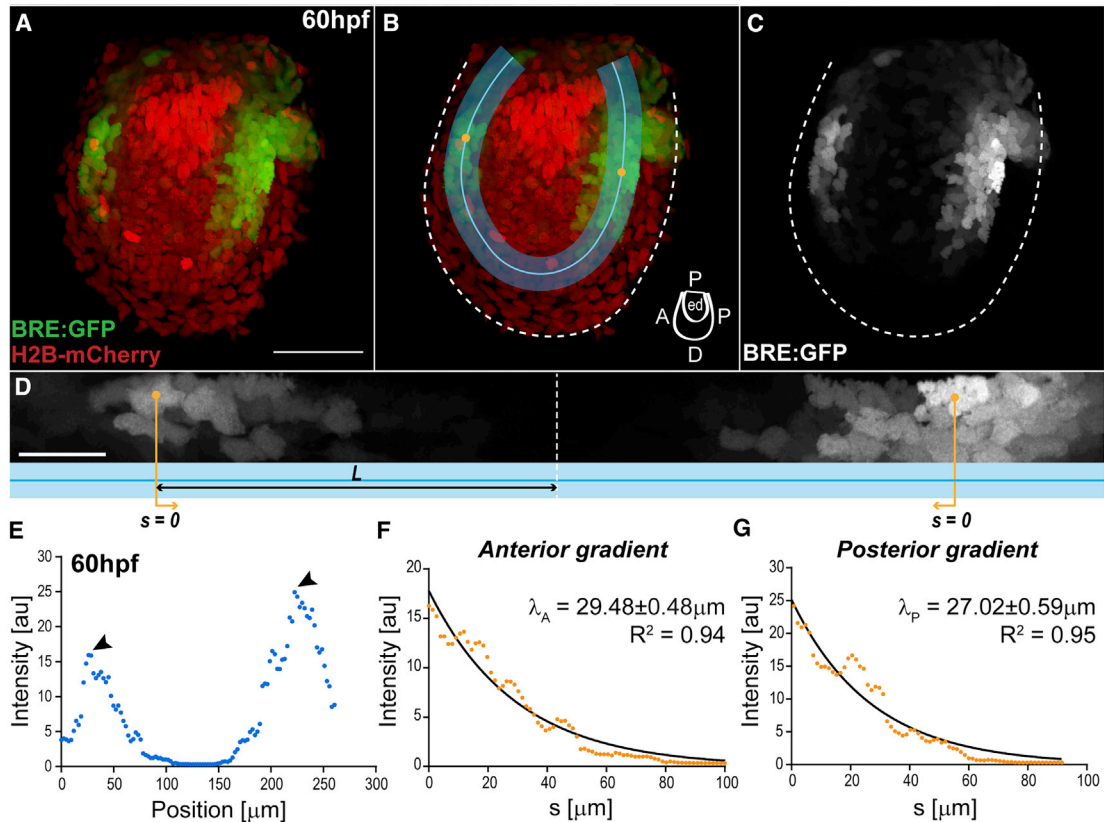
Figure 1. BMP Signaling Gradients in the Pectoral Fin

(A–C) Fin of double transgenic BRE:GFP (green) and Histone2b-mCherry (red), at 60 hpf. The BRE reporter expresses GFP under the control of Smad1/5/8 enhancers from the mouse Id1 promoter (Collery and Link, 2011; Laux et al., 2011). Dashed line, fin boundaries. Anterior, left; distal, down.

(B) Region of interest (ROI, blue; with ROI midline) where gradients are deployed: endoskeletal disc, abutting the fin fold. Cartoon indicates fin axes and endoskeletal disc (ed).

(D) BRE:GFP gradient along the straightened ROI (blue; anterior left) from (B). Orange lines indicate position $s = 0$, corresponding to intensity maxima. Length L (black line), distance between each peak signal and the ROI midpoint (dashed line).

Scale bars: 50 μm (A–C), 20 μm (D).

(E) BRE:GFP intensity profile along the ROI midline from (B)–(D), at 60 hpf. Arrowheads, intensity maxima. Intensity corresponds to signal average orthogonal to midline.

(F and G) Anterior (F) and posterior (G) intensity profiles from (E) versus position s, with respective decay lengths (λ, slope), ± SEM. Here $s = 0$, position of peak signal as indicated in (D)–(E). Note that the gradient profile corresponds to different levels of signaling per cell and the observed gradient is not due to different cell density or different number of signaling cells (see Figures S2K–S2M). Also, the gradient does not reflect different durations of signaling or long perdurance of GFP in the BRE reporter, since Phospho-Smad1/5/9 immunostainings show similar graded distributions (Figures S1J–S1L). Black lines, exponential fits with respective goodness of fit (R^2). BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011).

See also Figures S1 and S2.

S4K, S4L). We also studied the scaling of the gradient without committing to exponential fits, using an alternative method based on analyzing the collapse of the gradients into a single curve profile as the tissue grows (Figures S4A–S4D; Wartlick et al., 2011).

Because of homogeneous growth (Figures 2J–2L), the relative position of cells does not change as the tissue grows ($r_{cell}(t) = x(t)/L(t)$ is constant). In addition, due to scaling (Figures 3C and 3D), the normalized signaling level $C(r_{cell}, t)/C_0(t)$ in any particular cell in the tissue does not change over time (Aguilar-Hidalgo et al., 2018; Wartlick et al., 2011). Since $C(r_{cell}, t)/C_0(t)$ for each cell is constant, then the signaling level $C(r_{cell}, t)$ of each cell increases proportionally to the amplitude $C_0(t)$ of the gradient, i.e., when the amplitude of the gradient doubles in a particular cell, it doubles in all fin cells. Therefore, the dynamics of $C_0(t)$ at the source boundary reflect the dynamics of BMP signaling in every other cell in the tissue—this prompted us to study $C_0(t)$ in detail during pectoral fin growth. The amplitude C_0 increases as the fin grows, consistent with a power-law relationship $C_0 = L^q$, given the endoskeletal disc PD length L, with $q = 0.92 \pm 0.01$ (Figures 3E–3G). A value of the exponent close to one means effectively that the relationship between C_0 and L can be captured by a quasi-linear dependence. If gradient scaling fuels fin growth, what is the machinery of scaling?
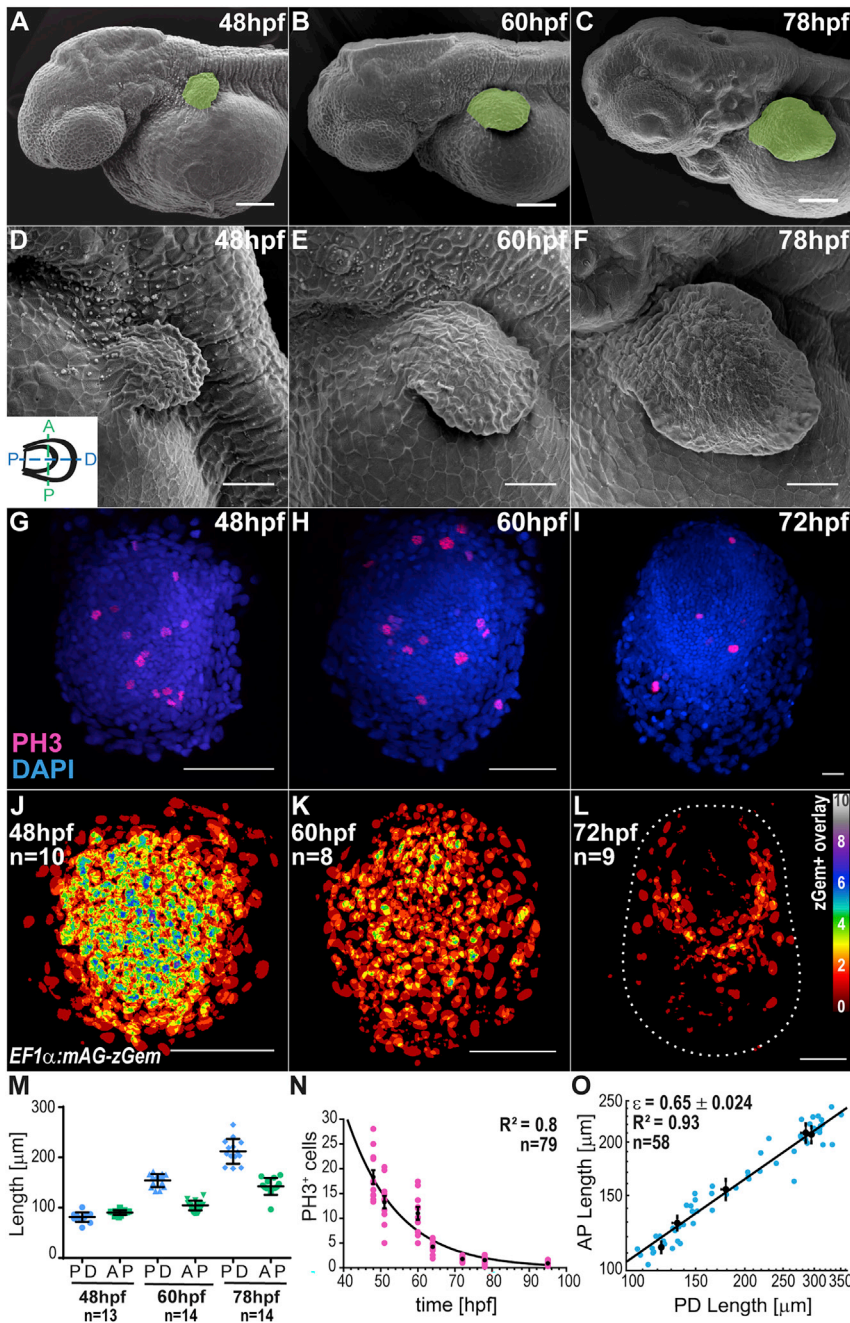
Figure 2. Pectoral Fin Growth

(A–F) Scanning electron micrographs of larvae (A–C) and fins (D–F) at different times. Cartoon indicates fin orientation.

(G–I) DAPI (blue) and Phospho-histoneH3 immunostainings (red) at different times.

(J–L) Overlay of fins expressing zGeminin. LUT shows number of overlaid zGeminin⁺ nuclei per pixel. Dashed line, fin boundary. Distal, down; anterior, left.

(M) PD and AP fin lengths from SEM images.

(N) Number of phospho-histoneH3 cells per 10,000 μm^2 at different times. Line, exponential fit with goodness of fit (R^2).

(O) Log-log plot of AP versus PD lengths to determine anisotropy ε (slope value) $\pm$ SEM. Line, power-law fit with goodness of fit (R^2).

Black dots, average from developmental time bins. Mean $\pm$ SEM are shown in all graphs. n represents total number of fins analyzed/overlaid. Scale bars, 100 μm (A–C), 50 μm (D–L).

Expander Molecule in Zebrafish: Smoc1

It has been proposed that secreted molecules can function as "expanders" of the gradient by acting on morphogen diffusion and/or degradation (Ben-Zvi and Barkai, 2010). These expanders would act as BMP agonists, extending the gradient. Beyond mere expansion, a feedback loop, where BMP signaling represses expander expression, can achieve scaling by keeping the decay length proportional to tissue length (Ben-Zvi and Barkai, 2010). This implies that the expander is expressed only at the edge of the gradient, i.e., at low levels of BMP signaling. Being restricted to the edge of the tissue, the expander expands

the gradient when the length of the gradient is shorter than the tissue itself: the expander adjusts (i.e., scales) the gradient to the tissue.

A *Drosophila* secreted molecule, Pentagone, fulfils these features and is essential to scale BMP2/4 gradients in the wing and eye (Vuilleumier et al., 2010; Wartlick et al., 2014). In heterologous overexpression experiments, Pentagone can cause BMP-like defects during dorso-ventral patterning in fish (Vuilleumier et al., 2010). In vertebrates, Sparc related modular calcium binding 1 (Smoc1) is the ortholog of Pentagone, and *Xenopus* Smoc1 has been shown to interfere with BMP signaling (Thomas et al., 2017). Furthermore, Smoc1 is associated with Waardenburg anophthalmia syndrome in humans and mutant mice show defects in their limbs and eyes, phenotypes associated with BMP defects (Abouzeid et al., 2011; Okada et al., 2011).

To study whether Smoc1 plays a role in BMP scaling in the fin, we analyzed the pattern of expression of Smoc1, as well as generated morphants and a CRISPR

mutant. Smoc1 is expressed as a gradient, with maximal concentration at the distal region of the fin, between the two BMP gradients, i.e., where BMP signaling levels are lowest (Figures 4A–4C, 4H, and S3A–S3K). This is consistent with the idea that BMP signaling itself represses Smoc1. Indeed, blocking BMP signaling by conditional expression of a dominant-negative version of the BMP receptor-1a (Pyati et al., 2005) upregulates Smoc1 (Figures 4D–4G and 4I), confirming the negative feedback loop. While Smoc1 expands the BMP gradient (Smoc1 agonizes BMP, see below), this negative feedback (BMP antagonizes Smoc1) closes a loop of regulation to achieve scaling, as
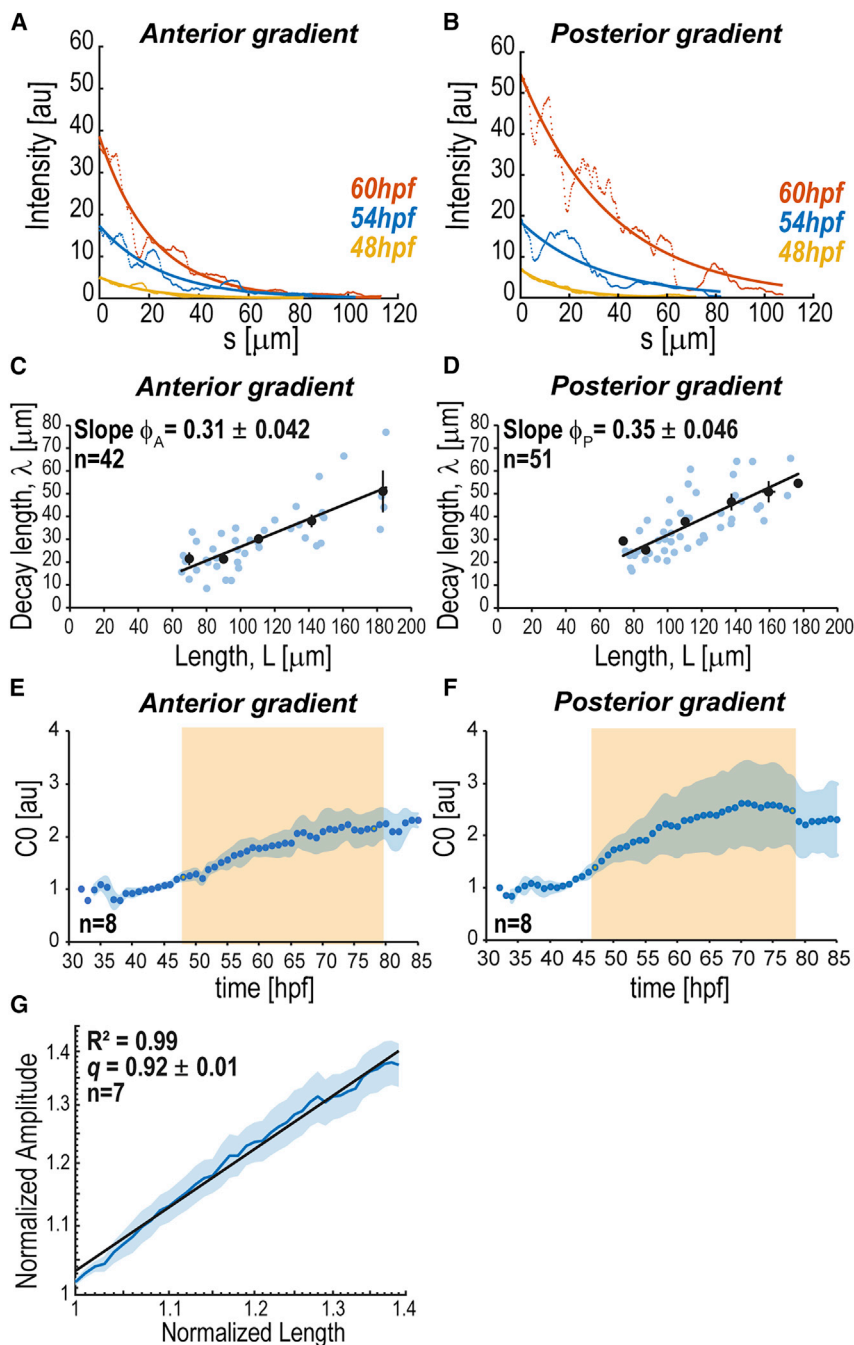
Figure 3. Scaling of BMP Signaling Gradients

(A and B) BRE:GFP gradients at different times as a function of the distance to intensity maxima ($s = 0$, ROI midline). Bold lines, exponential fits.

(C and D) Decay length versus ROI half-length (L) of BRE:GFP gradients. Blue dots, individual data; black dots, average from length bins ± SEM. Slope (ϕ) values ± SEM are shown. Lines, linear fits with goodness of fit: ϕ_A R^2 = 0.58; ϕ_P R^2 = 0.55.

(E and F) Average dynamics of gradient amplitude (C_0; SPIM imaging). Orange, growth phase (48–78 hpf).

(G) Log-log plot of normalized average amplitude $C_0(t)/C_0(45 \text{ hpf})$ versus normalized endoskeletal disc PD length $L(t)/L$ (45 hpf), to estimate q (slope value), with SEM. Power-law fit with goodness of fit (R^2).

Shadowing corresponds to SEM per time point (E, F) or length (G). n represents number of fins analyzed. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011) (A–D); BRE-AAVmlp:eGFP (Collery and Link, 2011) (E–G). See also Figures S2 and S4.

previously proposed in invertebrates (Ben-Zvi and Barkai, 2010; Vuilleumier et al., 2010).

Smoc1 Scales the Anterior Gradient and Controls Fin Growth

Is Smoc1 implicated in the scaling of the BMP gradient in the fin? To address this, we generated a Smoc1 CRISPR mutant that carries a stop codon in the 5′ region of the coding sequence (Smoc1^{ug104}; Figures 5A and S3M–S3U; STAR Methods). This mutant recapitulates a previously reported microphthalmia phenotype

caused by a validated splicing morpholino (Figures S3L and S3O–S3U; Abouzeid et al., 2011). Interestingly, *Smoc1* homozygous mutants fail to scale the anterior BRE gradient: the decay length does not expand as the fin length grows (Figures 5C, 5D, 5F, and S4W). Similar results were obtained with Phospho-Smad 1/5/9 immunostainings (Figures S4J, S4M, and S4N) and by performing collapse analysis (Figures 5L and S4E–S4H).

The specificity of the scaling phenotype is confirmed by the validated Smoc1 splicing morpholino (Figures S4Q–S4V; Abouzeid et al., 2011). Furthermore, upon expression of *smoc1* by mRNA injection into the CRISPR mutant background, the anterior gradient recovers its scaling, excluding a potential off-target effect (Figures 5F and S5; see also theoretical study of this experiment in Supplemental Methods).

Posterior gradient scaling is not affected in the mutant (Figures 5C, 5E, 5F, S4J, S4M, S4O, and S4X). In addition, we noticed that Smoc1 overexpression through mRNA injection in Smoc1 homozygous embryos led the two gradients to over scale, i.e., the scaling factors ϕ_L are larger than in wild type (Figures 5F, S5A, and S5B). The fact that the mutant interferes only with the anterior gradient but overexpression affects both gradients implies that the function of Smoc1 is redundant for the scaling of the posterior gradient. Indeed, it is possible that a second *smoc* gene (*smoc2*), present in chromosome 13, could play a key redundant role for posterior gradient scaling (Mommaerts et al., 2014).

Interestingly, loss of Smoc1 has consequences on fin growth. Figures 5G–5K show that growth is impaired in *Smoc1*
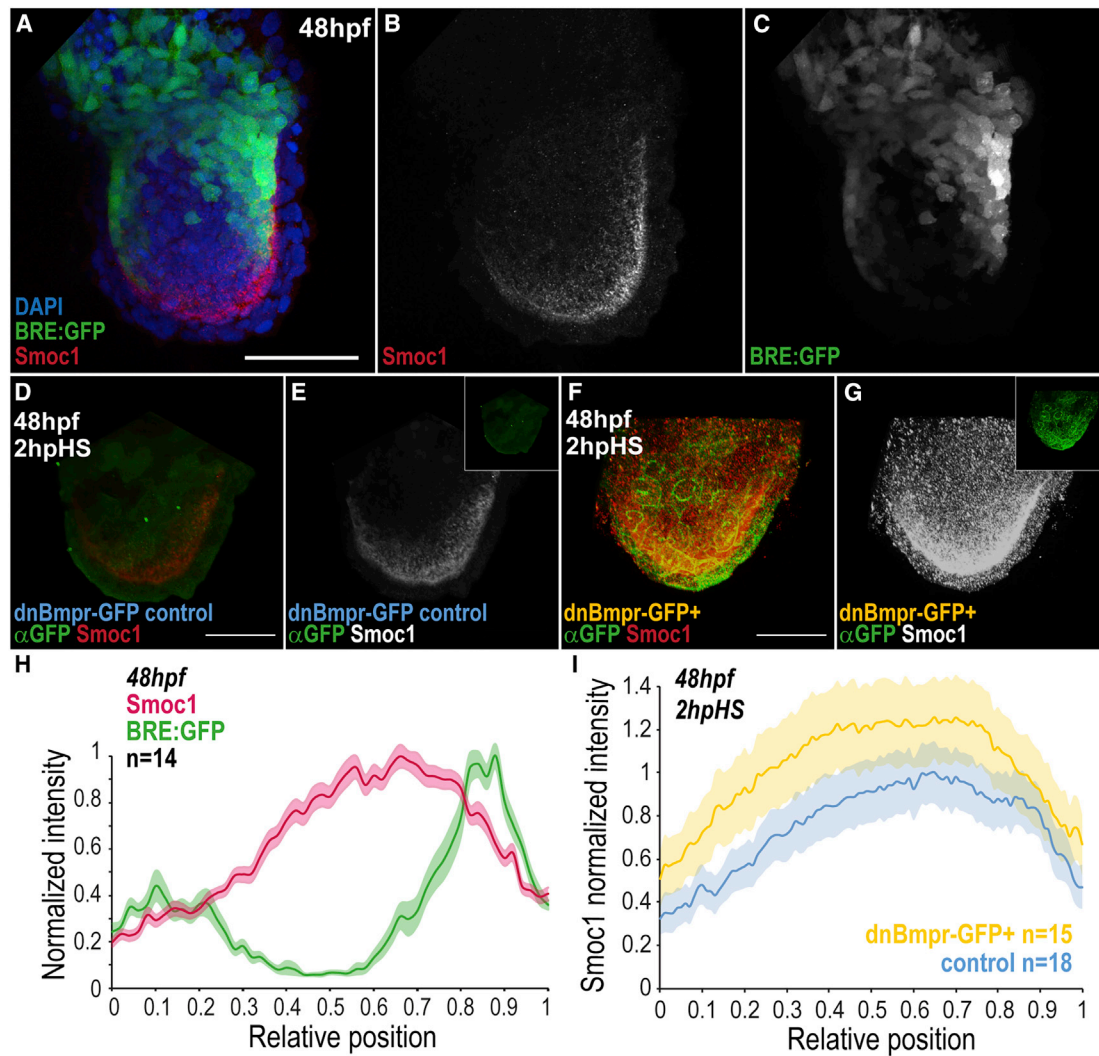
Figure 4. Smoc1 Is Repressed by BMP Signaling

(A–C) BRE:GFP signal (green) and Smoc1 immunostaining (red) in the fin, at 48 hpf.

(D–G) Smoc1 expression pattern at 2 h post-heat-shock induction (hpHS) of a dominant-negative BMP receptor transgene (dnBmpr-GFP[+]) or in sibling controls (for heat-shock conditions, see STAR Methods), at 48 hpf. Inset, GFP immunostaining to monitor expression of transgene.

(H) Average intensities of normalized Smoc1 immunostainings and BRE:GFP signal versus relative position (ROI midline), at 48 hpf. Intensities were normalized to respective maxima.

(I) Average intensity of normalized Smoc1 immunostainings versus relative position (ROI midline) in dnBmpr-GFP[+] transgenics and siblings control. Smoc1 intensity in dnBmpr-GFP[+] was normalized to the control maxima. Shadowing corresponds to SEM per relative position (H, I).

Scale bars, 50 μm. n represents number of fins analyzed. Anterior, left; distal, down. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011) (A–C, H). See also Figure S3.
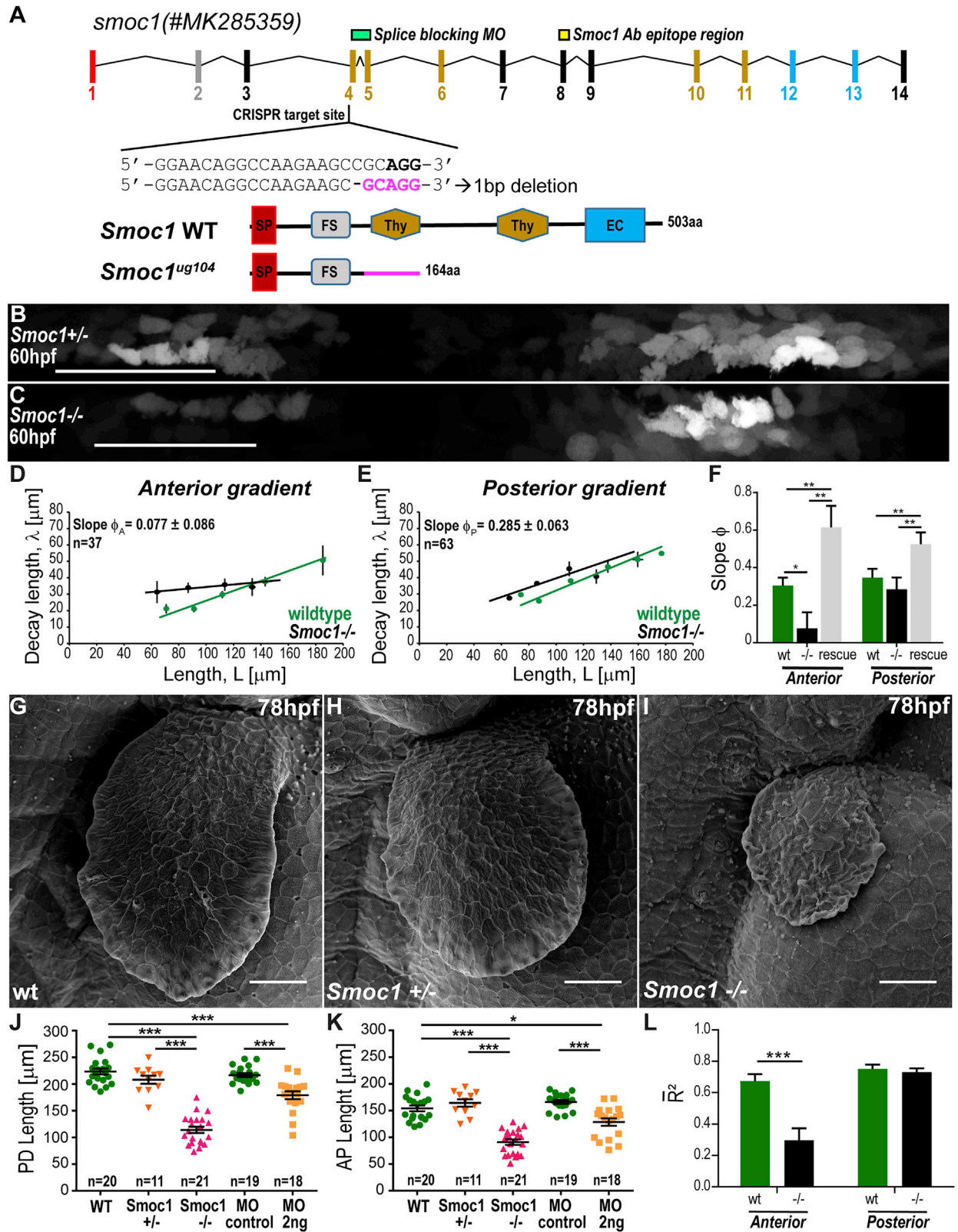
homozygote mutant and morphant fins (Figures S4Q and S4T). This is consistent with the idea that scaling of the gradient boosts the increase in BMP signaling levels necessary for growth, according to a growth mechanism based on the relative time derivatives of signaling.

DISCUSSION

Gradient Scaling in a Vertebrate Appendage

In this work, we show the scaling of two BMP signaling gradients (Figures 3C and 3D) that contribute to growth control of

the pectoral fin. A conserved secreted factor, Smoc1, mediates gradient scaling (Figures 5D and 5F) and itself regulates growth of this organ (Figures 5G–5K). Furthermore, the amplitude of the gradients display a power-law relationship with the size of the fin (Figure 3G). Gradient scaling and amplitude power-law are consistent with a process of growth control based on time derivatives of signaling, which can explain our observations where proliferation in the fin is homogeneous (Figures 2J and 2K). To give stronger support to this model, future experiments imposing different signaling derivatives (for instance, by manipulating BMP levels by heat-shock control at the single-cell level)

A

smoc1(#MK285359)

CRISPR target site

5'-GGAACAGGCCAAGAAGCCGC**AGG**-3'
5'-GGAACAGGCCAAGAAGC-GCAGG-3' → 1bp deletion

B
Smoc1+/-
60hpf

C
Smoc1-/-
60hpf

D *Anterior gradient*

Slope ϕ_A = 0.077 ± 0.086
n=37

wildtype
Smoc1-/-

E *Posterior gradient*

Slope ϕ_P = 0.285 ± 0.063
n=63

wildtype
Smoc1-/-

F

G 78hpf
wt

H 78hpf
Smoc1 +/-

I 78hpf
Smoc1 -/-

J

K

L

(legend on next page)

will address whether this changes the division rate of individual cells in the tissue.

In this report, we have mainly studied the BMP signaling gradients in the proximal-distal axis. Figures S1P–S1S, however, shows that these gradients are also deployed along the anterior-posterior axis, toward the center of the fin. Along the AP axis, the gradients' decay follows an exponential profile and their amplitude also increases over time: the dynamics of the gradients could in principle explain growth control by temporal derivatives of signaling in the entire fin primordium. Further experiments to analyze the gradients along the AP axis might reveal the logic of anisotropic growth in the fin. Indeed, our data show that gradient scaling influences the growth of the fin (Figures 5G–5K), consistent with such a mechanism based on temporal derivatives (Aguilar-Hidalgo et al., 2018; Wartlick et al., 2011, 2014). This mechanism has been seen in insects (Wartlick et al., 2011, 2014) and is seen now in a vertebrate, suggesting that this machinery could be widely employed throughout evolution.

Fin Growth and the Fold Change of Signaling Levels

We have previously shown in a context of scaling and homogeneous growth (Aguilar-Hidalgo et al., 2018; Wartlick et al., 2011, 2014) that a power-law relationship for C_0 with tissue size implies that the proliferation rate g_{cell} of each cell in the tissue is proportional to the normalized time derivative ($\dot{C}_{cell}/C_{cell}$) of the cellular level of BMP signaling, according to the equation $g_{cell} \cong ((1 + \varepsilon)/q)(\dot{C}_{cell}/C_{cell})$ (see Supplemental Methods for Theoretical Framework). The time derivative normalized to the actual levels of BMP signaling corresponds to the cellular fold-change of signaling over time. The observed gradient scaling and amplitude power-law imply that cell division occurs when the BMP signaling levels increase by 38% from the beginning of the cell cycle, considering the experimental values we observed for the anisotropy $\varepsilon = 0.65$ and the power-law coefficient $q = 0.92$ (for calculation, see Supplemental Methods).

Taken together, our data suggest a scenario where the fin contains a proximal BMP source that generates a single gradient deployed along the proximo-distal axis, while signaling, as revealed by the BRE reporter, is only seen in the ROI. This is consistent with the presence of BMP signaling antagonists in the endoskeletal disc (Bauer et al., 1998; Fürthauer et al., 1999), which could decrease the signaling levels in the center of the endoskeletal disc. This could generate an effective scenario where two

signaling gradients are deployed in the fin. Importantly, in this setting, the observed fin growth anisotropy (Figure 2O) and gradient scaling along the x axis (Figures S2D, S2E, S2G, and S2H) become compatible with the control of growth by relative increases of BMP signaling levels. Given that Smoc1 mutants affect only scaling of the anterior gradient, a Smoc1 paralog, Smoc2, might cover redundantly its function in the posterior gradient. It is possible that the BMP ligand(s) in the two gradients show different splicing forms, post-translational modifications, or binding partners whether the gene is expressed from the anterior or posterior source. One could speculate that this could allow for a differential mode of action for different Smocs in the anterior versus posterior sides of the fin. Local scaling within a single gradient could allow for local control of patterning and growth, which could give rise to the emergence of complex morphologies.

Computation of Time Derivatives of Signaling

Cell-to-cell communication mediated by fold change detection rather than absolute levels of signaling has been described in a number of sensory systems, from chemotaxis in bacteria and amoebas to olfactory or gustatory transduction in *C. elegans* (Adler and Alon, 2018; Bargmann, 2006; Barkai and Leibler, 1997; Frick et al., 2017; Friedrich and Jülicher, 2007; Levchenko and Iglesias, 2002). In particular, it has been proposed that within the TGFβ superfamily, not only BMP but also TGFβ signaling is mediated by cells measuring the relative increase in signaling over time (Frick et al., 2017; Sorre et al., 2014). In models where the interactome has been described (yeast and *E. coli*), networks that detect fold change in signaling are vastly over-represented (Alon, 2006; Mangan et al., 2006), suggesting that cells can compute more easily time derivatives than absolute concentration of molecules. In addition, a mechanism mediated by fold change detection and gradient scaling provides solutions supporting homogeneous cell proliferation and organ growth to a finite size (Aguilar-Hidalgo et al., 2018). Further exploration into the specific properties of a fold change detection system during the embryonic growth of the pectoral fin will provide a general framework to understand growth and patterning of vertebrate appendages.

STAR★METHODS

Detailed methods are provided in the online version of this paper and include the following:

Figure 5. Smoc1 Mutants Scaling Phenotype

(A) *smoc1* genomic structure. Exon colors match the protein domains they encode: SP, signal peptide; FS, follistatin-like domain; Thy, thyroglobulin-like domain; EC, extracellular domain. Target sequence of the *Smoc1^{ug104}* CRISPR mutant is located in exon 4; mutant sequence of *Smoc1^{ug104}*, highlighted in magenta. *Smoc1^{ug104}* results in a truncated protein, which lacks the first Thyroglobulin domain and other downstream features. The position of the splicing morpholino and Smoc1 antibody epitope region are indicated.

(B and C) BRE:GFP gradients in *Smoc1* homozygote mutants (C) and heterozygote controls (B) displayed as straightened ROI, as in Figure 1D (anterior, left).

(D and E) Decay length versus ROI half-length (*L*) of BRE:GFP gradients in *Smoc1*$^{-/-}$ mutants (black) and wild-type controls (green, from Figures 3C and 3D). Dots, average from length bins ± SEM (for individual data points, see Figures S4W–S4X). Lines, linear fits to individual data. Slope (ϕ) values ± SEM are shown.

(F) Comparison of $\phi = \lambda/L$ from BRE:GFP gradients of wild-type, *Smoc1*$^{-/-}$ mutants and *Smoc1*$^{-/-}$ mutants injected with *smoc1* mRNA ("rescue").

(G–I) Scanning electron micrographs of wild-type (G), *Smoc1* heterozygote control (H), and *Smoc1* homozygote fins (I).

(J and K) Average PD (J) and AP (K) fin lengths at 78 hpf in different mutant and morphant (MO) conditions.

(L) Comparison of average goodness of fit, R^2 ($\overline{R}^2$) obtained from collapse analysis (see Figures S4A–S4H) of BRE:GFP gradients in wild-type or *Smoc1*$^{-/-}$ mutants. n represents number of fins analyzed.

For all statistical analyses: ***p < 0.0001; **p < 0.01, *p < 0.05; two-tailed, unpaired, non-parametric Mann-Whitney tests. Mean ± SEM are shown in all graphs. Anterior, left; distal down. Scale bars, 50 μm. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011) (B–F, L). See also Figures S4 and S5.

SUPPLEMENTAL INFORMATION

Supplemental Information can be found online at https://doi.org/10.1016/j.celrep.2020.03.024.

AUTHOR CONTRIBUTIONS

R.M. performed most experiments and data analysis, with the help of C.S. and M.D. in biochemical experiments; L.H. built the SPIM microscope and performed associated experiments, as well as respective data analysis. C.S. cloned the updated *smoc1* cDNA. R.M. established the $Smoc1^{ug104}$ mutant line; R.M., L.H., and Z.H. wrote MATLAB and ImageJ/Fiji code for data analysis. R.M., L.H., and M.G.-G. conceived and designed experiments; F.J. developed the 2D theory of growth control by time derivatives of signaling and Z.H. the scenario of homogeneous expander; R.M. and M.G.G. prepared the manuscript.

DECLARATION OF INTERESTS

The authors declare no competing interests.

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STAR★METHODS

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
rabbit anti-BmpR1ba+b	Genetex	Cat# GTX128200; RRID: AB_2827925
rat anti-PhosphoH3 (pSer28)	Sigma	Cat# H9908; RRID: AB_260096
mouse anti-Smoc1	Abnova	Cat# H00064093-M03; RRID: AB_530203
rabbit anti-PhosphoSmad 1 (Ser463/465) /5 (Ser463/465) /9 (Ser465/467)	Cell Signaling	Cat# 13820; RRID: AB_2493181
rabbit anti-GFP	Abcam	Cat# ab290; RRID: AB_303395
rat anti-GFP	Sta Cruz Biotechnology	Cat# sc-101536; RRID: AB_1124404
rabbit anti-mCherry	Living Colors	Cat# 632496, lot# 1408015; RRID: AB_10013483
rabbit anti-panActin	Thermofisher	Cat #MA5-11869; RRID: AB_11004139
Chemicals, Peptides, and Recombinant Proteins		
smoc1 splice blocking morpholino	Gene Tools	Abouzeid et al., 2011
smoc1 mismatch control morpholino	Gene Tools	This paper
DMH1	Sigma	CAS# 1206711-16-1
MS-222	Sigma	CAS# 886-86-2
Deposited Data		
smoc1 cDNA sequence	This study	GenBank: MK285359
Experimental Models: Organisms/Strains		
Zebrafish: AB wildtype strain	European Zebrafish Resource Center	RRID: ZDB-GENO-960809-7
Zebrafish: Tg(EF1α:mAG-zGem(1/100))[rw0410 h]	Sugiyama et al., 2009	RRID: ZDB-ALT-101018-2
Zebrafish: Tg(Xla.Eef1a1:hist2 h2l-mCherry)	Recher et al., 2013	RRID: ZDB-FISH-150901-2743
Zebrafish: Tg(BmpRE:EGFP)[pt509]	Laux et al., 2011	RRID: ZDB-ALT-110706-57
Zebrafish: Tg(BRE-AAVmlp:eGFP)[mw29]	Collery and Link, 2011	RRID: ZDB-ALT-110308-1
Zebrafish: Tg(BRE-AAVmlp:d2EGFP)[mw30]	Collery and Link, 2011	RRID: ZDB-ALT-110310-1
Zebrafish: Tg(hsp70l:dnXla.Bmpr1a-GFP)[w30]	Pyati et al., 2005	RRID: ZDB-ALT-050503-2
Zebrafish: *Smoc1*[ug104]	This study	RRID: ZDB-ALT-180625-1
Oligonucleotides		
smoc1 cDNA:FWD: 5′-GATCGGCGCGCCCATGAACTGTCACAATCTGGC-3′	This study	N/A
smoc1 cDNA: REV: 5′-CGCTGGCCGCCGACCTTACTGAACAAATTTG-3′	This study	N/A
smoc1[ug104] cDNA:FWD: 5′-CAGGCCAAGAAGCGCAGGAATCCATTTTC-3′	This study	N/A
smoc1[ug104] cDNA: REV: 5′-GAAAATGGATTCCTGCGCTTCTTGGCCTG-3′	This study	N/A
smoc1[ug104] sgRNA:FWD: 5′-TAGGAACAGGCCAAGAAGCCGC-3′	This study	N/A
smoc1[ug104] sgRNA: REV: 5′-AAACGCGGCTTCTTGGCCTGTT-3′	This study	N/A
Smoc1[ug104] genotyping primers:FWD: 5′-CAAGAATCCATTCCCCCC-3′	This study	N/A
Smoc1[ug104] genotyping primers: REV: 5′-CTTTAGGATTCGTTGATGTCAAAAG-3′	This study	N/A
Splicing block MO flanking region primers:FWD: 5′-CTTGTTTCAGAGGTCAAAGAGTTG-3′	Abouzeid et al., 2011	N/A

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Continued

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Splicing block MO flanking region primers: REV: 5′-GATTAAGCAGCTGGTGTACAAAGAG-3′	Abouzeid et al., 2011	N/A
Recombinant DNA		
pCS2-smoc1 for mRNA and *in situ* probe synthesis	This study	N/A
pCS2-smoc1-mCherry for mRNA synthesis	This study	N/A
pCS2-smoc1^{ug104}-mCherry for mRNA synthesis	This study	N/A
pCS2-nls-zCas9-nls for mRNA synthesis	Jao et al., 2013	N/A
pDR274-smoc1sgRNA for CRISPR generation	This study	N/A
Software and Algorithms		
MATLAB R2018b	Mathworks	https://www.mathworks.com/products/matlab.html
Imaris 8.0	Bitplane	https://imaris.oxinst.com/packages
Fiji	Schindelin et al., 2012	https://fiji.sc
Prism 7	GraphPad	http://www.graphpad.com/scientific-software/prism
Ilastik 1.3.0	Haubold et al., 2016	https://www.ilastik.org/
Lasergene 15	DnaStar	https://www.dnastar.com/software/lasergene/
FinchTV	Geospiza	https://www.geospiza.com/finchtv
Excel	Microsoft	https://products.office.com/en-us/?rtc=1
ZiFiT	Sander et al., 2010	http://zifit.partners.org/

LEAD CONTACT AND MATERIALS AVAILABILITY

Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Marcos Gonzalez-Gaitan (marcos.gonzalez@unige.ch). Plasmids and zebrafish lines generated in this study are available upon request.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Ethics statement
This study followed European Union directives (2010/63/EU), the Swiss Animal Protection Act and the Swiss Animal Welfare Ordinance.

Zebrafish lines and maintenance
All zebrafish (*Danio rerio*) lines used were maintained in a re-circulating system with a 14 h/day, 10 h/night cycle at 28°C. Crosses were performed with 3- to 12-month old wildtype AB strain adults. Embryos were kept in E3 zebrafish embryo medium at 28.5°C until reaching the desired developmental stage. The transgenic lines used were: Tg(EF1α:mAG-zGem(1/100))$^{rw0410\ h}$ (Sugiyama et al., 2009) (RRID: ZDB-ALT-101018-2); Tg(Xla.Eef1a1:hist2 h2l-mCherry) (Recher et al., 2013) (RRID: ZDB-FISH-150901-2743) – referred as H2B-mCherry; Tg(BmpRE:EGFP)^{pt509} (Laux et al., 2011) (RRID: ZDB-ALT-110706-57), Tg(BRE-AAVmlp:eGFP)^{mw29} (Collery and Link, 2011) (RRID: ZDB-ALT-110308-1) - for simplicity both are denominated BRE:GFP, each figure legend details which transgene was used; Tg(BRE- AAVmlp:d2GFP)^{mw30}, referred as BRE:d2GFP (Collery and Link, 2011) (RRID: ZDB-ALT-110310-1); Tg(hsp70l:dnXla.Bmpr1a-GFP)^{w30} (Pyati et al., 2005) (RRID: ZDB-ALT-050503-2), referred as dnBmpr-GFP.

METHOD DETAILS

SPARC related modular calcium binding 1 (Smoc1) sequence and constructs
The Ensembl sequence for *Danio rerio smoc1* (ENSDARG00000088255) as it appears in the repository misses a signal peptide. NCBI BLAST was performed to identify a predicted signal peptide sequence in the current genome assembly (GRCz11). In addition, an extra exon at the end of the sequence (exon 14) was identified by homology to *Sinocyclocheilus rhinocerous smoc1* sequence (taxID: 307959). This was followed by amplifying the updated *smoc1* cDNA from 5 days post-fertilization (dpf) zebrafish total cDNA together with primers containing AscI and FseI digestion sites at their 5′ and 3′ ends, respectively (FWD: 5′-GATCGGCGCGCCCATGAACTGTCACAATCTGGC-3′; REV: 5′-CGCTGGCCGCCGACC TTACTGAACAAATTTG-3′). The 1530bp amplified fragment was purified using a gel extraction kit (QIAGEN 28706), digested with restriction enzymes AscI and FseI (NEB) and cloned into pCS2+8CmCherry (Addgene #34935) – which was previously digested with the same restriction enzymes and treated with Calf intestinal alkaline phosphatase (NEB M0290S). The updated *smoc1* cDNA sequence has been deposited into GenBank: MK285359.

To obtain the construct pCS2-smoc1^{ug104}-mCherry, which recapitulates the *Smoc1^{ug104}* fish CRISPR mutation, the pCS2-smoc1-mCherry construct was mutated by site directed mutagenesis, using PCR. For this, 600ng of template pCS2-smoc1-mCherry and primers containing the mutation, overlapping the mutation target site were used:

FWD 5′-CAGGCCAAGAAGCGCAGGAATCCATTTTC-3′;

REV 5′-GAAAATGGATTCCTGCGCTTCTTGGCCTG-3′.

Then 1μL of DpnI (NEB) was added directly to the PCR reaction to digest the template DNA at 37°C o/n. This was followed by transformation of competent cells, antibiotic selection, DNA amplification and sequencing for identification of the correct clone.

Smoc1^{ug104} CRISPR mutant

Smoc1^{ug104} CRISPR mutants (RRID: ZDB-ALT-180625-1), were generated as described (Hwang et al., 2013). Briefly, oligos for sgRNA synthesis were designed using ZiFiT software (Sander et al., 2010), (FWD: 5′-TAGGAACAGGCCAAGAAGCCGC-3′; REV: 5′-AAACGCGGCTTCTTGGCCTGTT-3′) annealed and ligated into the pDR274 expression vector (previously digested with BsaI, NEB). Then the sgRNA was synthetized using the Megascript T7 kit following manufacturer's protocol (Ambion). sgRNA purification was performed with the Mega Clear kit (Invitrogen), following its elution procedure. In parallel, Cas9 mRNA was synthetized from pCS2-nls-zCas9-nls plasmid (Jao et al., 2013). One-cell stage wildtype AB embryos were injected with 1nl of a mix containing 300pg Cas9 mRNA and 150pg sgRNA. To test efficiency of sgRNA, injected embryos were genotyped and T7 endonuclease I (NEB) mutation detection assays were performed, as well as sequencing of the region of interest. Founders were identified by genotyping the resulting progeny. From F2 generation onward, *Smoc1^{ug104}* mutants were maintained in the BRE:GFP; H2B-mCherry background. For rescue experiments, 20pg of full-length *smoc1* mRNA were injected at one-cell stage into homozygote *Smoc1^{ug104/ug104}*; BRE:GFP; H2B-mCherry embryos and imaged from 48 to 78 hpf. For synthesis details, see below.

Microinjection of embryos with mRNAs and morpholinos

One-cell stage wildtype AB or *Smoc1^{ug104}*; BRE:GFP; H2B-mCherry embryos were injected using standard procedures with different *smoc1* mRNA concentrations. This was generated by linearization of *pCS2-smoc1* vector with NotI (NEB), and transcribed using the SP6 mMESSAGE mMACHINE High Yield Capped RNA Transcription Kit (Ambion). The same procedure was performed to obtain *smoc1-mCherry*, *smoc1^{ug104}-mCherry* and Cas9 mRNAs. For antisense morpholino oligonucleotide (MO) injections, 2ng or 12ng of a splicing block MO (5′-CCGGAACTCTGACAGACCTGAGCAA-3′) (Abouzeid et al., 2011) and its respective mismatch control MO (5′-CCGCAAGTCTCACACACCTCAGCAA-3′) were used (Gene Tools, LLC) in BRE:GFP; H2B-mCherry heterozygote embryos. Embryos were left to develop at 28.5°C until the desired stage and then processed for SEM, live imaging or immunostainings. A PV-820 Pico-injector (World Precision Instruments) and a Narashige micromanipulator were used for microinjection.

Genotyping

Genomic DNA from embryos or adult caudal fins was extracted using the Kapa Express Extract kit (Kapa Biosystems) according to the manufacturer's protocol. This was followed by performing PCR with KAPA2G Robust HotStart ReadyMix (Kapa Biosystems) with primers surrounding the *smoc1* mutation region (FWD: 5′-CAAGAATCCATTCCCCCC-3′; REV: 5′-CTTTAGGATTCGTTGATGTC AAAAG-3′). Then PCR products were purified using a PCR Purification Kit (QIAGEN) and sent to sequencing. Sequences were analyzed using Finch TV software.

Embryo heat-shocks

For heat-shocks, 48 hpf embryos collected from outcrossing heterozygote hsp70:dnBmpr1-GFP with wildtype AB fish were placed into 50ml falcons with pre-heated E3 medium and heat activated in a bath at 38.5°C for 1 h, then transferred to Petri dishes and kept at 28.5°C. GFP positive and negative embryos were selected and fixed for immunostainings 2 h after heat-shock. To assess fin growth upon inhibition of BMP signaling with hsp70:dnBmpr1-GFP, the same procedure as above was performed in intervals of 12 h from 48 until 72 hpf, when embryos were selected and fixed for SEM.

DMH1 treatment

For inhibition of BMP signaling with DMH1 (Hao et al., 2010), embryos collected from in-crossing heterozygote BRE:d2GFP; H2B-mCherry transgenic fish were used. At 30 hpf, embryos were selected for both GFP and mCherry signals, dechorionated and placed in 6-well culture dishes (Falcon) containing either DMH1 50μM (Sigma) or DMSO control (Sigma) diluted in E3 embryo medium. Embryos were kept at 28.5°C until 72 hpf, when embryos were live-imaged. Chemical treatments were repeated 3 times with different biological replicates.

Total RNA isolation and Reverse Transcriptase (RT)-PCR

Total RNA was extracted from 5 dpf embryos using Trizol reagent (Invitrogen) and treated with DnaseI (Roche) according to the manufacturer's protocol. cDNA was synthesized from 1 μg total RNA using the SuperScript IV First-Strand Synthesis System (ThermoFisher), following the oligo-dT manufacturer protocol.

For morpholino splicing block test (Figure S3L), total RNA extraction and cDNA synthesis from 24 hpf injected embryos was performed as described above. Then a PCR using primers flanking the splicing region (FWD: 5′-CTTGTTTCAGAGGTCAAAGA GTTG-3′, REV: 5′-GATTAAGCAGCTGGTGTACAAAGAG-3′) was performed and visualized by 1% agarose gel electrophoresis.

Immunofluorescence

Whole-mount immunostainings were performed as described (Mateus et al., 2012). Embryos were mounted in 80% Glycerol, 2% DABCO (Sigma) diluted in PBS and then imaged using a Zeiss LSM780 confocal microscope with a C-Apochromat 40 × water objective or a C-Apochromat 20 × dry objective. The primary antibodies used were: rabbit anti-BmpR1b (BmpR1ba+b 1:200, Genetex); rat anti-PH3 (1:300, Sigma), mouse anti-Smoc1 (1:200, Abnova), rabbit anti-PSmad1/5/9 (1:100, Cell Signaling), rabbit anti-GFP (1:200, Abcam), rat anti-GFP (1:200, Sta Cruz Biotechnology) and rabbit anti-mCherry (1:200, Living Colors #632496). The secondary antibodies used were: anti-mouse Alexa488, anti-mouse Alexa594, anti-mouse Alexa647, anti-rat Alexa488, anti-rat Alexa594, anti-rabbit Alexa488, anti-rabbit Alexa594 (all 1:500, Molecular probes). Immunostainings were repeated at least 3 times with different biological replicates, per marker and condition.

In situ hybridization

Whole-mount *in situ* hybridizations in embryos were performed as described (Thisse and Thisse, 2008). After histochemical reaction, embryos were fixed in 4% paraformaldehyde, dehydrated progressively into 100% methanol, cleared with benzyl alcohol:benzyl benzoate:xylene mix and mounted in Permount (Fisher, #10271893)(Bauer et al., 1998). Images of *in situ* hybridizations were obtained with a Leica M80 stereomicroscope, using a Leica IC80 HD digital camera. Digoxygenin-labeled antisense RNA probe for *smoc1* was synthesized as described (Henrique et al., 1995), by linearization of smoc1-pgem-T-easy with AatII (NEB) and transcribed with SP6 polymerase (Promega). *In situs* were repeated 3 times with different biological replicates.

Western blot

For protein extracts, 6 hpf wildtype embryos non-injected or injected with either 130pg *smoc1-mCherry*, 130pg *smoc1^{ug104}-mCherry* or 130pg of *smoc1* mRNA were dechorionated by adding 1ml Pronase (10 mg/mL, Sigma) directly to the respective Petri dish, incubating 5 min at RT and followed by washing 3 times with E3. Then 130 embryos of each condition were transferred to 1.5 mL eppendorf tubes and 1 mL of deyolking buffer (55mM NaCl, 1.8mM KCl, 1.25mM $NaHCO_3$) was added. This was followed by shaking for 5 min at 1100rpm, centrifugation at 300 g for 1 min and removing the supernatant. Then embryos were washed with wash buffer (110mM NaCl, 3.5mM KCl, 2.7mM $CaCl_2$, 10mM TrisHCl pH 8.5) and the centrifugation repeated, leading to a pellet of cells. Cell pellets were subsequently flash frozen in liquid nitrogen and stored at −80°C. Afterward, cell pellets were squashed with a pestle into 100μl of lysis buffer (50mM Tris pH 7.5, 1.5mM $MgCl_2$, 125mM NaCl, 0.2% IGEPAL, 5% glycerol and protease inhibitor cocktail – benzamidine (1mM, Applichem), chymostatin (40μg.ml^{-1}, Applichem), antipain (40μg.ml^{-1} Applichem), leupeptine (1μM Applichem), pefabloc (1mM) and PMSF (0.5mM)). The extracts were incubated 40cmin at 4°C with rocking, then cellular debris were cleared by centrifugation at 16000 g for 10cmin at 4°C. The concentration of the extracts was measured using Bradford (Biorad 500-0205) and the extracts were normalized to the least concentrated sample. They were then diluted in LDS sample buffer (Life Technologies) enriched with 2.5% β-mercaptoethanol and analyzed by SDS–PAGE and western blot. In each condition, 20μg of total protein were loaded. SDS–PAGE was performed using NuPAGE 4%–12% Bis-Tris gels (Life Technologies) according to the manufacturer's instructions. Gels were transferred to nitrocellulose membranes using iBLOT (Life Technologies) according to the manufacturer's instructions. For western blot, the antibodies used were: rabbit anti-mCherry (1:1000, Living Colors #632496), mouse anti-Smoc1 (1:1000, Abnova) and rabbit anti-panActin (1:10000, Thermofisher MA5-11869). Antibodies were diluted in TBS with 0.2% BSA, 1 mM $CaCl_2$, 0.02% Thymerosal O/N at 4°C. Western blots were revealed using HRP coupled antibodies (Jackson immunoResearch 1:10000 dilution), Western Bright Quantum (Advansta) chemiluminescence reagents and a Vilber Lourmat Fusion imager. Western blots were repeated 3 times with different biological replicates. For gel source data, see Data S1.

Scanning electron microscopy (SEM)

Larvae were fixed o/n at 4°C in 2.5% glutaraldehyde, 4% paraformadehyde, 0.1 M Cacodylate buffer, 2 mM Calcium Chloride. Then larvae were washed in 0.1 M Cacodylate buffer and fixed a second time in 1% OsO_4 in 0.1 M Cadodylate buffer, for at least 2 h on ice. After this, larvae were washed in H_2O and subsequently dehydrated to 100% EtOH, in 10% steps. Then larvae were transferred to 100% acetone and kept at 4°C until imaging. Before imaging, larvae went through critical point processing for total dehydration using a Leica EM-CPD030. Then embryos were mounted, gold coated in a JEOL JFC-1200 fine coater and imaged using a JEOL JSM-6510LV. Mounting was performed so that the fin apical surface was parallel to the imaging plane. SEM was repeated at least 3 times with different biological replicates.

Live imaging

Confocal live imaging was performed in embryos anaesthetized in 0.1% MS-222 (Sigma) diluted in embryo medium, using a Fluorodish (World Precision Instruments), and an inverted Zeiss LSM780 confocal microscope with a C-Apochromat 40 × water objective.

Light-sheet microscopy was performed using a home-built Selective Plane Illumination Microscopy (SPIM), based on the original design (Huisken et al., 2004), with adaptations. A cylindrical lens with a focal length $f = 50$ mm (Thor Labs GmbH) was used to generate the light sheet and focused onto the sample using a 10x, NA 0.3 water-immersion objective (UMPLFLN10XW, Olympus). Two lasers were used (λ = 488 nm, Oxxius Laserboxx, and λ = 561 nm, Cobolt Jive) to excite two fluorophores at the same time. An Acousto-Optical Tunable Filter (AOTF, AA OptoElectronic) was used to select the wavelength, intensity and duration of the excitation light. Imaging was done using a 20x, NA 0.5 water-immersion objective (UMPLFLN20XW, Olympus), a $f = 150$ mm 2" achromatic lens (Thor Labs GmbH), a dichroic mirror (t560lpxr, Chroma Technology) to split the fluorescence emission and two cameras (Photometrics CoolSNAP HQ2). Zebrafish larvae were mounted as described (Kaufmann et al., 2012), using a motorized XYZ-rotation stage (Physik Instrumente). Live imaging was repeated at least 5 times with different biological replicates per condition and zebrafish line.

QUANTIFICATION AND STATISTICAL ANALYSIS

Proliferation
Quantification of the proliferation patterns and rates were performed using a Fiji (Schindelin et al., 2012) plugin, ObjectJ (Figures 2G–2I and 2N). This plugin was used to manually identify all PH3 positive nuclei and to measure the respective fin area, to obtain the mitotic density in the fin. The mitotic density data was fitted to a temporal exponential decay and its goodness of fit (R^2) assessed.

For proliferation density maps (Figures 2J–2L), we first projected the z stack images of fins expressing zGeminin. We then segmented the zGeminin positive nuclei using Ilastik software (Haubold et al., 2016). From each developmental time point, all segmented images were sum-overlaid in Fiji, registering spatially the fins by using their centroid and the PD/AP axes. This way, we record for each pixel the number of fins in which at least one zGeminin positive nuclei is present. We used a lookup table to display the number of overlaid zGeminin positive nuclei per pixel, per n fins. The total number of fins was similar between the different developmental stages considered (n = 8 to 10).

Fin dimensions and length comparison
The AP and PD fin lengths were manually measured by drawing a segmented line in SEM images or live imaging images (BRE:GFP; H2B-mCherry transgenics), using Fiji (Figures 2A–2F and 2M). For the PD axis, the most proximal point was chosen by eye, while the most distal point was defined as the most distant point from the proximal point, at the edge of the fin. The AP axis was defined to be perpendicular to the PD axis, approximately at its midpoint, to capture the widest region of the fin. To measure endoskeletal disc PD or DV length in the SPIM images, the same procedure as above was applied in maximum intensity projections from the H2B-mCherry signal.

Fin length statistical analysis was performed using GraphPad Prism software by carrying out non-parametric, unpaired, two-tailed Mann Whitney t tests between the different conditions. Statistical experimental details can be found in figure legends of Figures 5J, 5K, S1C, and S1D.

Anisotropy
Fin anisotropy ε (Figure 2O) was quantified from the PD and AP length values (L_{PD}, L_{AP}) from the live imaging dataset. Anisotropy is defined as the ratio of the growth rates g along the AP versus PD axis $\varepsilon = (g_{AP}/g_{PD})$, where $g_{AP} = (\dot{L}_{AP}/L_{AP})$ with $\dot{L}_{AP}$ the time derivative of the L_{AP} length and $g_{PD} = (\dot{L}_{PD}/L_{PD})$. Therefore $(\dot{L}_{AP}/L_{AP}) = \varepsilon(\dot{L}_{PD}/L_{PD})$ and taking integrals $log(L_{AP}) = \varepsilon \log(L_{PD}) + C$, where C is an integration constant. Thus, as in previous reports (Aguilar-Hidalgo et al., 2018; Wartlick et al., 2014), we estimated anisotropy by fitting to the length data the power-law relationship $L_{AP} \sim L_{PD}^{\varepsilon}$ and displaying the relationship as a log-log plot where the slope of the line corresponds to the anisotropy.

BRE:GFP Gradient Quantification
After image acquisition, we performed 3D rotation, cropping and flattening by using IMARIS software so that the fin endoskeletal disc was aligned along the AP/PD plane. Maximum intensity projections (MIP) were obtained and a region of interest was defined using Fiji (Figures 1B and 1D). The region of interest (ROI) where the gradient is deployed was defined as an area of 50 pixels width in the endoskeletal disc, along the region abutting the apical fold, which was identified by nuclear density monitored with the H2B-mCherry signal. The BRE:GFP signal intensity in a particular position along the ROI midline corresponds to the average signal in their orthogonal positions within the ROI. The two positions with maximal intensity along the ROI midline were identified as the positions of amplitude C_0 of the anterior and posterior gradients. No background subtraction was performed in the analysis of BRE:GFP intensity profiles.

In Figures 1F, 1G, 3A–3D, 5D, 5E, S4K–S4P, S4R–S4S, S4U–S4X, S5A, and S5B the position s in each of the two gradients is defined as the distance to the corresponding peak intensity (C_0 position) following the path along the ROI midline. Using this coordinate system, system length L was defined as the distance between each peak signal and the ROI midpoint (ROI half-length). Alternatively, in Figures S2D–S2J, the position x corresponds to the distance to the signal peak along the PD axis and the position y, to the distance along the AP axis.

In any of these two cases, gradient profiles were fitted to an exponential curve to determine their decay length (λ), goodness of fit (R^2) and confidence values, with custom-made MATLAB code. In order to obtain decay length estimations of good quality, we

discarded gradient fits with $R^2 < 0.8$ (see Data S2 for statistical analysis to determine this threshold and full datasets without threshold exclusion). In scaling plots, the selected decay lengths were represented versus fin length and fit to linear regressions. Note that, for this regression, an offset was allowed. In conditions in which there is scaling, the slope (scaling factor, ϕ) departs significantly from zero and this offset is negligible. For instance, in Figures 3C and 3D, the offsets were very small (-3.79 ± 4.95 μm and -2.83 ± 5.29 μm for anterior and posterior wild-type gradients, respectively), so we neglected them. In contrast, when scaling fails (Figures 5C, S4E, and S4G), the slope is closer to zero and the offset is not negligible (26.743 ± 8.82 μm and 19.32 ± 19.18 μm, respectively).

The rationale behind the choice of fitting to a linear regression with or without an offset is as follows. In wild type, the decay length λ is proportional to the length of the fin L. This means that there is no offset: the ratio λ/L is constant. If there is an offset, this is not true, but instead $(\lambda + \text{offset})/L = \text{constant}$. When there is "$\lambda/L = \text{constant}$" scaling, then growth control by $\dot{C}/C$ generates homogeneous growth (see Wartlick et al., 2011), but not in the other case, when there is an offset. The actual data confirm what is expected: there is no offset in wild type. We allowed an offset when doing the fits in wild type, verifying that these offsets were extremely small (see Figure S4P). When scaling fails (in mutants), there is an "intrinsic length scale," and this intrinsic length of the gradient should not change as the tissue grows. Indeed, in the mutant, there is an offset (the intrinsic decay length of the gradient in the absence of the expander) and this does not change over time, i.e., the slope of the regression is close to zero. This scenario is what we observe experimentally (see Figure S4P). In Figures 3C, 3D, 5D, 5E, S4K, S4L, S4N, and S4O, data were also averaged within length bins and the SEM is given.

Phospho-Smad1/5/9 Gradient Quantification

Quantification of PSmad1/5/9 gradient profiles and scaling plots (Figures S1J–S1L, S4I–S4L, S4N, and S4O) were performed as described for BRE:GFP above. To remove high background levels resulting from the immunostaining, individual z stacks were processed with Fiji plugin Remove Outliers (for bright outliers with radius of 2.0 pixels; threshold 50) and individual intensity profiles were subtracted by the respective minimum intensity value, before fitting to exponential curves to determine their decay length (λ), goodness of fit (R^2) and confidence values.

Quantification of the spatial intensity profile of other markers (H2B-mCherry, Smoc1, Smoc1-mCherry, BmpR1b, BRE:d2GFP) along the ROI midline was performed as described above. Individual profiles were normalized by dividing intensity and length values (entire ROI length) by respective maxima (unless stated in figure legends), followed by interpolation, averaging and plotting mean together with SEM.

Statistical analysis of ϕ and offsets

Statistical comparison of gradient scaling factors, ϕ (in Figure 5F for BRE:GFP gradients, Figure S4M for PSmad1/5/9 gradients) and offsets (Figure S4P) obtained from linear fits of scaling plots, was done using GraphPad Prism software by performing non-parametric, unpaired, two-tailed Mann Whitney t tests between the different conditions. Statistical experimental details can be found in respective figure legends.

Gradient collapses, density and statistical analysis of $\overline{R}^2$

To study scaling of the BRE gradients from the raw data without fitting exponentials, we performed analysis of the gradient profiles $C(r,t)$, where $r = x/L$ is the relative distance to the C_0 position. This showed that the relative concentration gradient $C(r,t)/C_0(t)$ remains unchanged during growth of the fin. The normalized $C(r,t)/C_0(t)$ profiles "collapse" into a time- and size-independent master curve $\xi(r) = (C(r,t)/C_0(t))$.

For gradient collapses (Figures S4A, S4C, S4E, and S4G), gradient intensity measurements obtained previously were used. Here, full datasets were considered – without performing exclusions based on goodness of exponential fit (R^2). Like before, the maximum intensity peaks of anterior and posterior gradients were identified as well as their respective length. Taking into account that gradients have different lengths, individual normalization by dividing intensity and length values by respective maxima were performed, followed by interpolation and plotting. An average curve and respective standard deviation were then calculated and plotted. To determine the goodness of the collapse, each individual gradient was fit to the average curve (without assuming an exponential profile) and respective R^2 was calculated. $\overline{R}^2$ corresponds to the average of all obtained R^2.

Statistical comparison of $\overline{R}^2$ obtained in the different conditions was done using GraphPad Prism software by performing non-parametric, unpaired, two-tailed Mann Whitney t tests between the different conditions. Statistical experimental details can be found in figure legends of Figure 5L.

Gradient density plots (Figures S4B, S4D, S4F, and S4H) of the obtained collapses were obtained by using a lookup table where each color corresponds to the fraction of normalized gradients passing through a certain bin. Bin size: 0.05 for ordinates and 0.05 for abscissas.

AP gradients

A region of interest (AP ROI) was defined using Fiji on the maximum intensity projections (MIP) obtained from BRE:GFP images above. The AP ROI where the anterior-posterior gradient is deployed was defined as an area of 90 pixels height in the endoskeletal disc, covering the two positions with maximal BRE:GFP intensity in the anterior and posterior fin sides (Figure S1P–S1R). From there analysis proceeded as for BRE:GFP gradients above.

Gradient signaling analysis per cell

Intensity from BRE:GFP signal in individual cells was performed in straightened posterior gradient ROIs (Figure S2K). From there, proximal (at the source of maximum BRE intensity) and distal (at the end of the visible BRE signal) ROIs were defined. Then, three individual cells per region (per gradient) were outlined in Fiji and intensity measured per cell. To quantify the number of BRE positive cells in gradients (Figure S2L), the same proximal and distal regions were used and ROIs were defined (113px width/height, approximately 20 cells), to allow for nuclei identification (H2B-mCherry signal) and BRE+ counting.

Statistical comparison was done using GraphPad Prism software by performing non-parametric, unpaired, two-tailed Mann Whitney t tests between the different conditions. Statistical experimental details can be found in figure legends.

Image processing of SPIM movies

SPIM movies (typically > 200 Gb) originally stored in MicroManager TIFF-files were converted to ImageJ-TIFF files containing one time point per file. All stacks were cropped in three dimensions to the same size. Registration of time points was done on a 3x downscaled version of the H2B-mCherry dataset to reduce RAM usage and processing time. Nuclei from the H2B-mCherry channel were used as reference points for the Descriptor Based Registration Plugin (Preibisch et al., 2010), assuming that the average movement of all nuclei between adjacent time points is close to zero. The obtained 3D translation matrix was then applied to the original dataset using MATLAB. The resulting movie was cropped in 3D and the registration procedure was repeated if necessary. The final translation was then applied to the BRE:GFP channel as well. For 3D visualization purposes and C_0 measurements, image processing ended here.

Morphogen gradient quantification of SPIM movies

After image acquisition and processing, the fin PD- and AP-axis were aligned to the *xy* plane, each twentieth time point per movie. For this, three points were manually selected in the pectoral fin: a distal point close to the PD axis, an anterior point close to the AP axis and a posterior point close to the AP axis. Then, an affine transformation matrix to translate/rotate these points into the *xy* plane was calculated and the transformation was performed using the TransformJ affine transform plugin (Meijering et al., 2001). After this, a rotation around the z axis was performed to align the PD axis to the y axis. Finally, the image stack was cropped in *z*. To apply these transformations to intermediate time points, linear interpolation was performed to calculate the three points, consequently calculating the respective affine transformation matrix to intermediate time points. Similar interpolation was applied to obtain the rotation around the z axis and the z-crop for intermediate time points. These transformations were then applied on the whole movie using Fiji. At this point, a maximum intensity projection was obtained from the resulting movie. Final registration was done on the MIP of the BRE:GFP channel using the Descriptor Based Registration Plugin (Preibisch et al., 2010) allowing for Rigid 2D transformations.

To measure BRE:GFP gradients over time, a similar procedure as above for ROI establishment and intensity measurement was performed on the MIP of every fifth image. For the intermediate time points, the spatial coordinates of the line were linearly interpolated and the intensity profile for these time points was measured based on the interpolated line.

Gradient scaling was measured using a model-free intensity-based fitting approach, using MATLAB. Time-adjacent intensity profiles were fit to each other with four fit parameters: Length-scaling factor (S_L, range: 0.8-1.2); Intensity-scaling factor (S_I, range: 0.8-1.2); Length offset (O_L, range: ± 10 μm) to compensate for length differences caused by manually drawing the line; and Intensity offset (O_I, range: 200-500) to include a background signal that does not scale. The value of O_I within this range did not significantly influence the values for S_L, S_I and O_L. To prevent a halt of the fitting optimizer in a local intensity minimum, we used a range (equally spaced points within the range of each fitting parameter) of initial estimates as input to the optimizer and chose the one with the lowest error.

Normalized C_0 was calculated by taking the cumulative product of BRE:GFP S_L for each time-point (between 45 and 72 hpf), while setting S_L = 1 for T = 0. Ultimately two curves are generated in this way, one in the forward-time direction (fit the curve at T = t+1 to the curve at T = t) and one in the reverse-time direction. The final curve is the average of those two curves.

To determine q (Figure 3G), the individual normalized C_0 values were plotted against the respective fin endoskeletal PD length normalized to its length at 45 hpf in log-log scale. Data averaging and respective SEM were then calculated and a power-law fit to the average was performed to determine the slope value q.

Measuring C_0 (Figures 3E and 3F): For each time point, C_0 was measured independently from the normalized C_0 by obtaining the maximum value in a manually selected rectangular/circular region around the location with the highest intensity in the MIP using Fiji (Schindelin et al., 2012) (Video S1).

DATA AND CODE AVAILABILITY

Source data for Figures 1, 2, 3, 4, 5, and S1–S5 are available in the online version of the paper. The accession number for the updated *smoc1* cDNA sequence reported in this paper is GenBank: MK285359. Custom Fiji and MATLAB code to analyze gradients are available upon request.
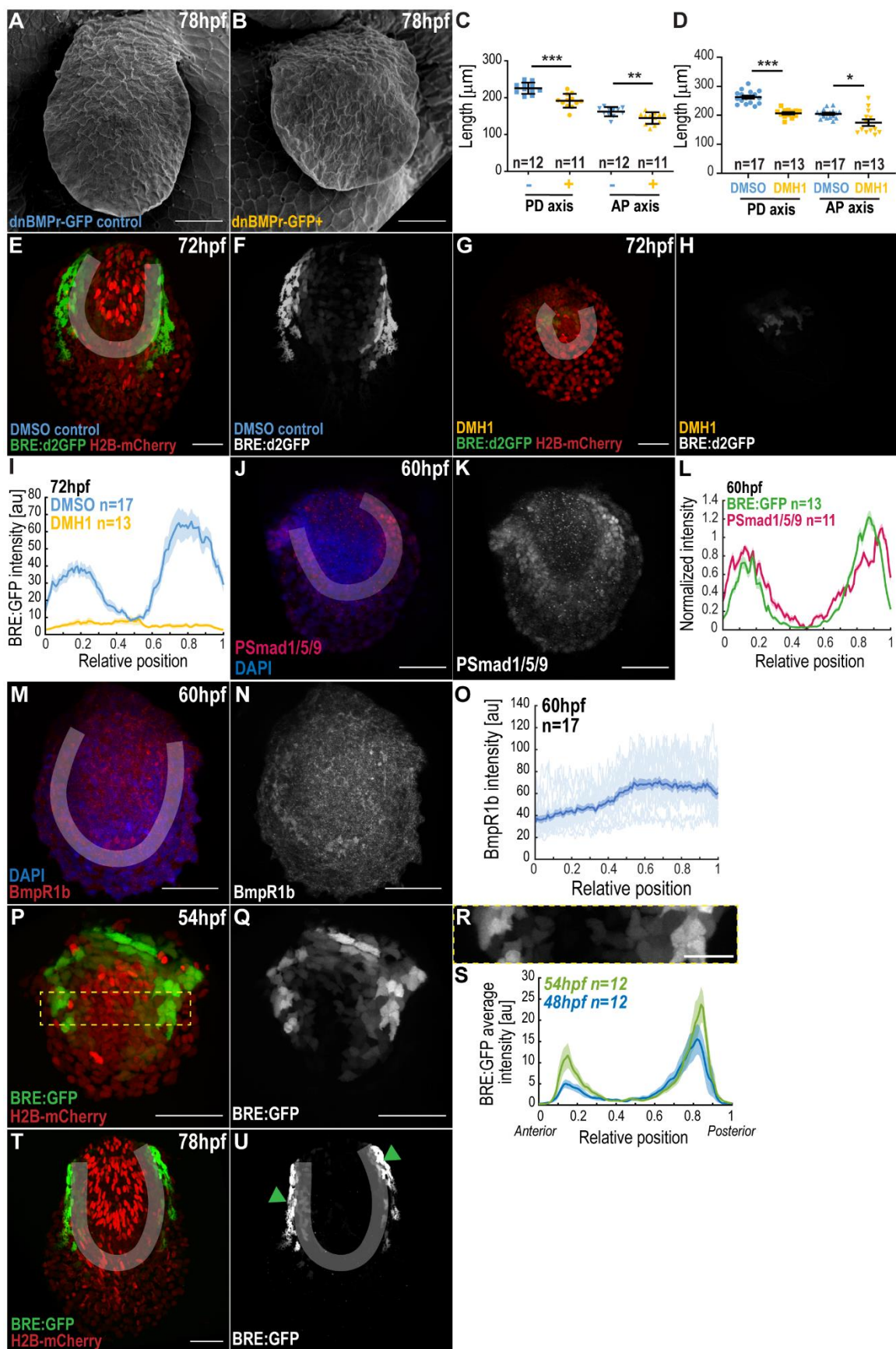
Supplemental Information

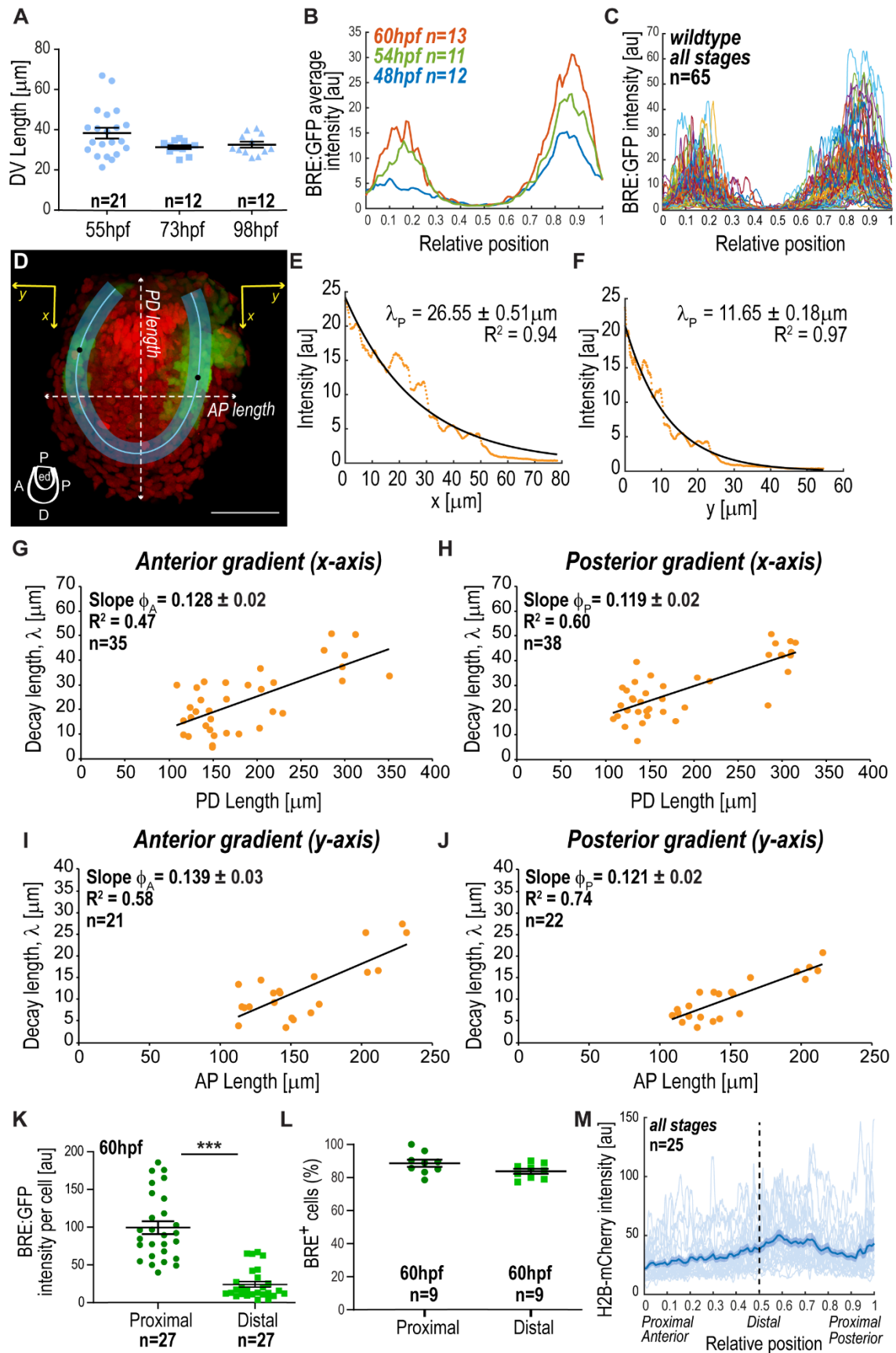**BMP Signaling Gradient Scaling**

in the Zebrafish Pectoral Fin

Rita Mateus, Laurent Holtzer, Carole Seum, Zena Hadjivasiliou, Marine Dubois, Frank Jülicher, and Marcos Gonzalez-Gaitan

Supplemental Figures

Figure S1. Characterization of BRE:GFP signalling gradients. Related to Figure 1.
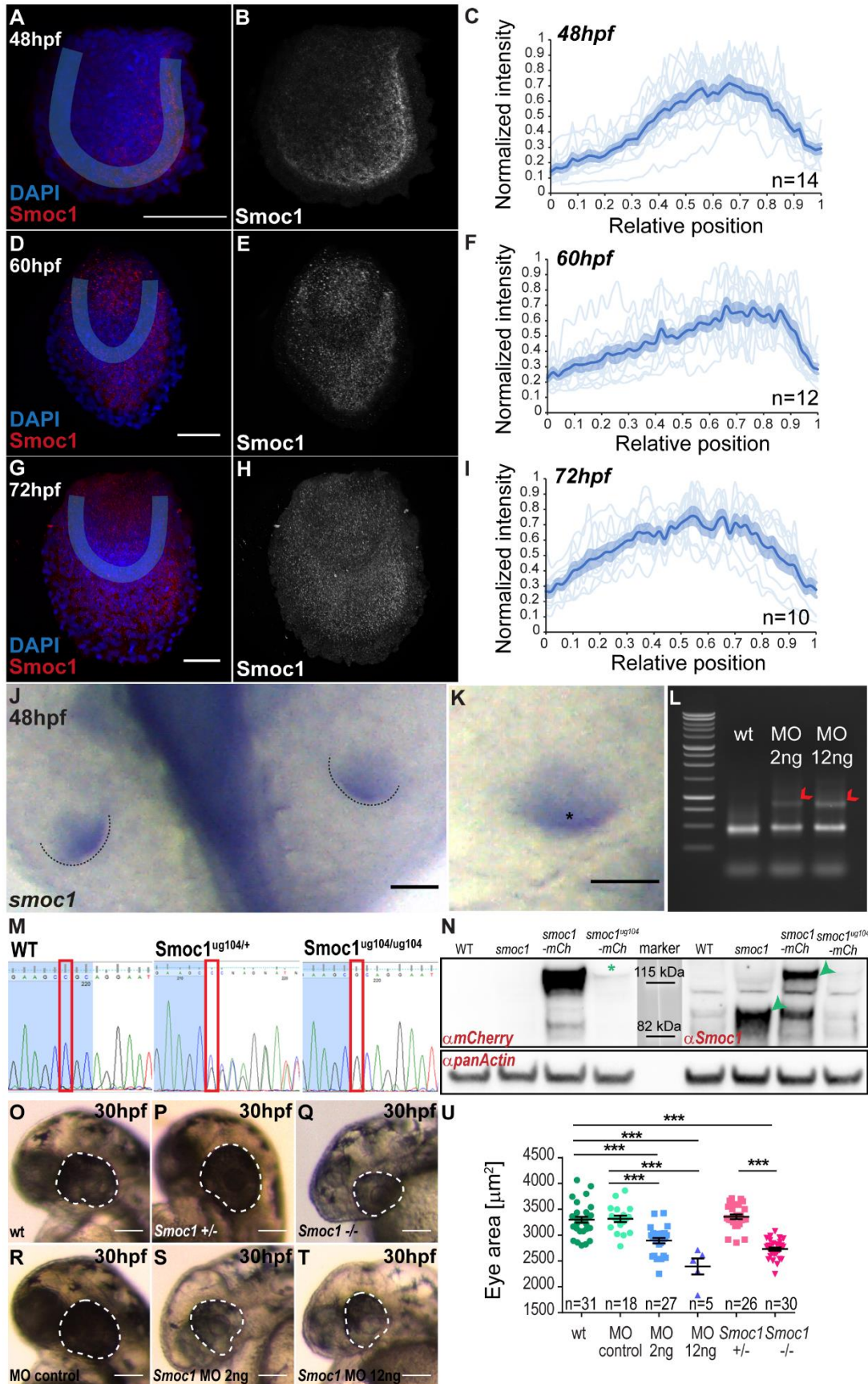
A-I Inhibition of BMP signalling leads to fin growth defects. **A-B** SEM of fins expressing dominant negative BMP receptor in transgenic fish (dnBmpr-GFP+, B) or in sibling controls (A), at 78 hpf (for heat-shock conditions, see STAR Methods). **C** Average PD and AP lengths of fins in conditions as in A-B. **D** Average PD and AP lengths of fins of conditions as in E-H. Mean ± s.e.m. are shown in all bar graphs. For all statistical analyses: ***P<0.0001; **P<0.01; *P<0.05; two-tailed, unpaired, non-parametric Mann–Whitney tests. **E-H** Live imaging of BRE:d2GFP; H2B-mCherry transgenics, in DMH1-treated (G-H) and control larvae (E-F), at 72 hpf (for inhibitor conditions, see STAR Methods). **I** Average intensity of BRE:GFP signal *versus* relative position (ROI midline) in DMH1-treated (yellow) and control larvae (blue), in conditions as in E-H. **J-K** Phospho-Smad1/5/9 immunostaining (red) and DAPI (blue), at 60 hpf. **L** Comparison between average intensities of normalized Phospho-Smad1/5/9 (red, from immunostaining) and BRE:GFP (green, from live imaging) versus relative position (ROI midline). **M-N** Bmp receptor 1b immunostaining (red) and DAPI (blue), at 60 hpf. **O** Average intensity of BmpR1b signal versus relative position (ROI midline). **P-Q** Fin of double transgenic BRE:GFP (green) and Histone2b-mCherry (red) at 54 hpf. Dash outline delimits the AP ROI used for quantification. **R** Cropped BRE:GFP gradient along the AP ROI from P. **S** Average intensity of BRE:GFP signal *versus* relative position (AP ROI), at 48 and 54 hpf. **T-U** Live imaging of BRE:GFP (green); H2B-mCherry transgenics (red), at 78 hpf. Note that at 78 hpf (and later, not shown) BRE:GFP drives high levels of expression in individual cells (arrowheads) in the fin fold adjacent to the endoskeletal disc in a pattern different to the gradient considered in this study. All shadowing in graphs, s.e.m. per relative position. n represents number of fins analysed. All fin images: anterior, left; distal, down. Region of interest (ROI midline), white highlight. BRE:GFP transgene used: BRE:d2GFP (Collery and Link, 2011) (E-I); BRE:eGFP (Laux et al., 2011) (L-P, T-U). Scale bars, 50µm (A-B, E,G,J-K, M-N, P-R, T); 20µm (R).

A

DV Length [μm]

n=21 — 55hpf
n=12 — 73hpf
n=12 — 98hpf

B

BRE:GFP average intensity [au]

60hpf n=13
54hpf n=11
48hpf n=12

Relative position

C

BRE:GFP intensity [au]

wildtype all stages n=65

Relative position

D

y *x* *y* *x*

PD length

AP length

P A ed P D

E

Intensity [au]

$\lambda_P = 26.55 \pm 0.51\mu m$
$R^2 = 0.94$

x [μm]

F

Intensity [au]

$\lambda_P = 11.65 \pm 0.18\mu m$
$R^2 = 0.97$

y [μm]

G

Anterior gradient (x-axis)

Decay length, λ [μm]

Slope ϕ_A = 0.128 ± 0.02
R^2 = 0.47
n=35

PD Length [μm]

H

Posterior gradient (x-axis)

Decay length, λ [μm]

Slope ϕ_P = 0.119 ± 0.02
R^2 = 0.60
n=38

PD Length [μm]

I

Anterior gradient (y-axis)

Decay length, λ [μm]

Slope ϕ_A= 0.139 ± 0.03
R^2 = 0.58
n=21

AP Length [μm]

J

Posterior gradient (y-axis)

Decay length, λ [μm]

Slope ϕ_P= 0.121 ± 0.02
R^2 = 0.74
n=22

AP Length [μm]

K

BRE:GFP intensity per cell [au]

60hpf

***

Proximal n=27   Distal n=27

L

BRE⁺ cells (%)

60hpf n=9 — Proximal
60hpf n=9 — Distal

M

H2B-mCherry intensity [au]

all stages n=25

Proximal Anterior   *Distal*   *Proximal Posterior*

Relative position

Figure S2. Cartesian system to measure the gradients and scaling. Related to Figures 1 and 3.
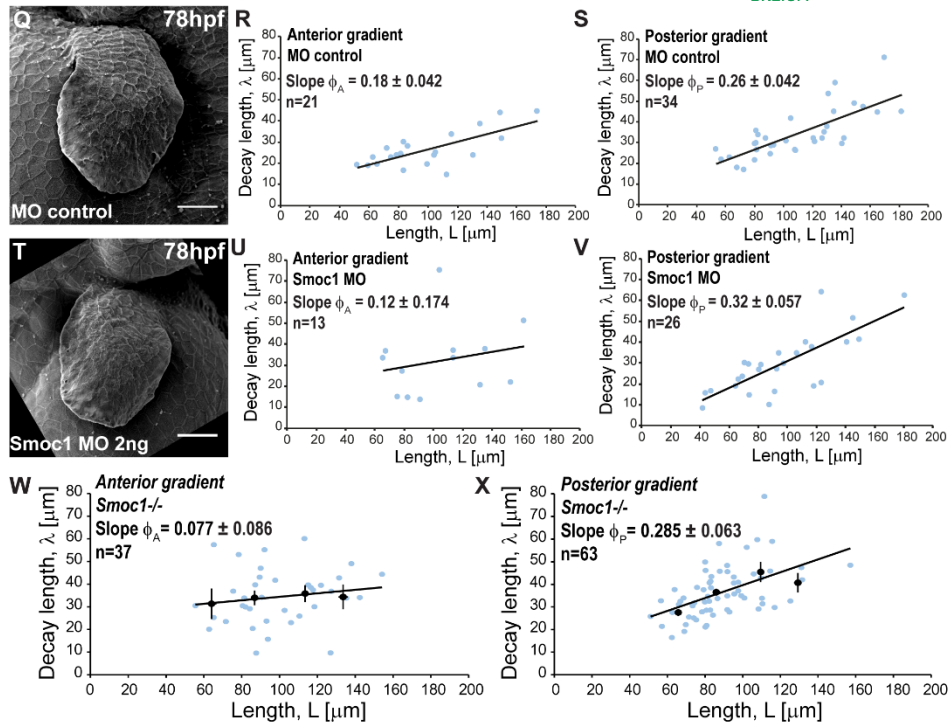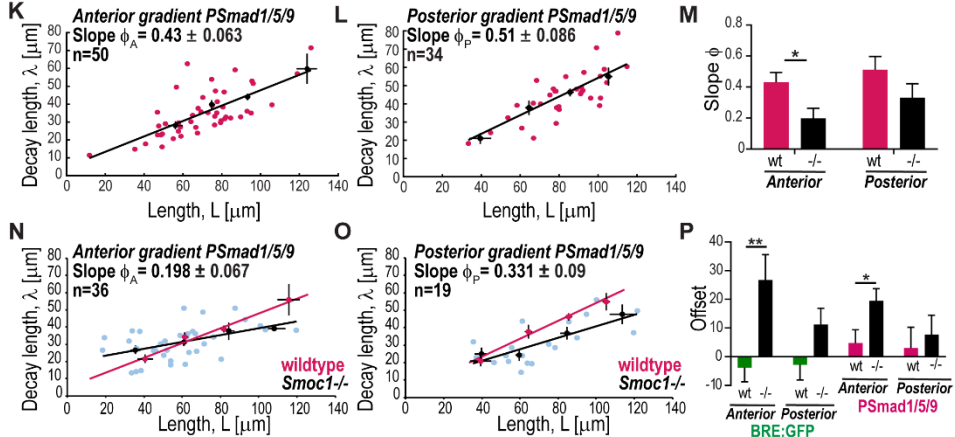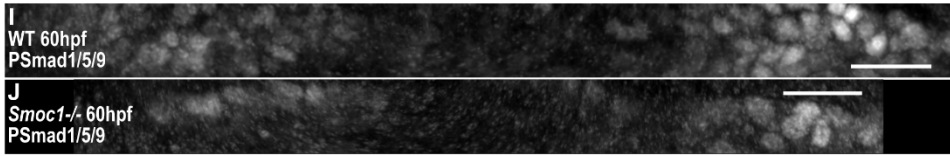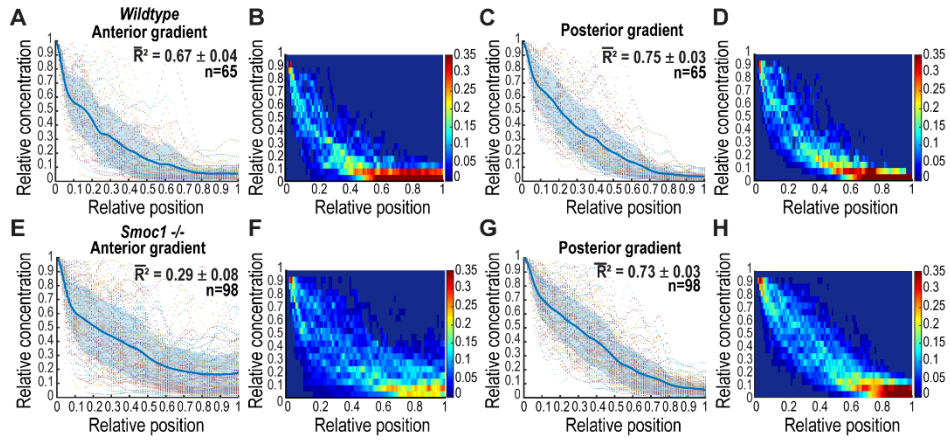
A DV fin lengths (from live imaging of H2B-mCherry transgenics), at different developmental stages. Note that there is no growth along the DV axis in the stages considered in this study. **B** Average intensities of BRE:GFP at different developmental stages *versus* relative position (ROI midline). **C** Intensity of individual BRE:GFP profiles *versus* relative position (ROI midline) from 48 to 78 hpf. **D** Fin of double transgenic BRE:GFP (green) and Histone2b-mCherry (red) from Fig. 1A. Region of interest (ROI), with midline (blue). PD length, distance between the proximal base of the fin and its distal tip along the PD axis. AP length, distance between the anterior-most part of the fin to its posterior-most position (white lines). Cartoon indicates fin axes and endoskeletal disc (ed). Scale bar, 50μm. **E-F** BRE:GFP intensity profiles versus x (E) or y (F) position for the posterior gradient in D, with respective decay lengths (λ, slope), ± s.e.m. Here $x = 0$ (E) or $y = 0$ (F), position of peak signal as indicated in D (black dots). Black lines, exponential fits with respective goodness of fit (R^2). **G-J** Decay lengths λ from exponential fits obtained from BRE:GFP intensity profiles in *x versus* PD length (G-H) or in *y versus* AP length (I-J). Slope (ϕ) values ± s.e.m. are shown. Lines, linear fits with goodness of fit (R^2). **K-M** Gradient signalling levels decrease along ROI midline, at 60 hpf. **K** Comparison between BRE:GFP average intensity per cell in proximal and distal areas of the gradient. **L** Comparison between the percentage of H2B-mCherry positive cells which are positive for BRE:GFP in proximal and distal areas of the gradient. **M** Nuclear density measured by H2B-mCherry signal along the ROI midline, showing that cell density is approximately homogeneous along the endoskeletal disc. Light blue, individual fin profiles; dark blue line, average intensity; shadowing, s.e.m. per relative position. Black dash indicates the most distal position in ROI, where density is similar to proximal positions. For all statistical analyses: ***P<0.0001; two-tailed, unpaired, non-parametric Mann–Whitney tests. Mean ± s.e.m. are shown in all graphs. n represents number of fins analysed, except in K, where n equals number of cells. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011).

A 48hpf DAPI Smoc1
B Smoc1
C 48hpf n=14
D 60hpf DAPI Smoc1
E Smoc1
F 60hpf n=12
G 72hpf DAPI Smoc1
H Smoc1
I 72hpf n=10
J 48hpf smoc1
K *
L wt MO 2ng MO 12ng
M WT Smoc1ug104/+ Smoc1ug104/ug104
N WT smoc1 smoc1-mCh smoc1ug104-mCh marker WT smoc1 smoc1-mCh smoc1ug104-mCh 115 kDa 82 kDa α.mCherry α.Smoc1 α.panActin
O wt 30hpf
P Smoc1 +/- 30hpf
Q Smoc1 -/- 30hpf
R MO control 30hpf
S Smoc1 MO 2ng 30hpf
T Smoc1 MO 12ng 30hpf
U Eye area [μm²] *** *** *** *** *** n=31 n=18 n=27 n=5 n=26 n=30 wt MO control MO 2ng MO 12ng Smoc1 +/- Smoc1 -/-
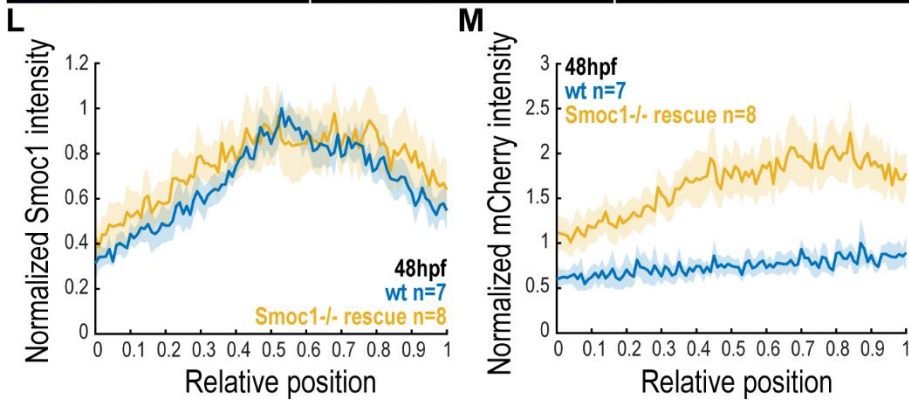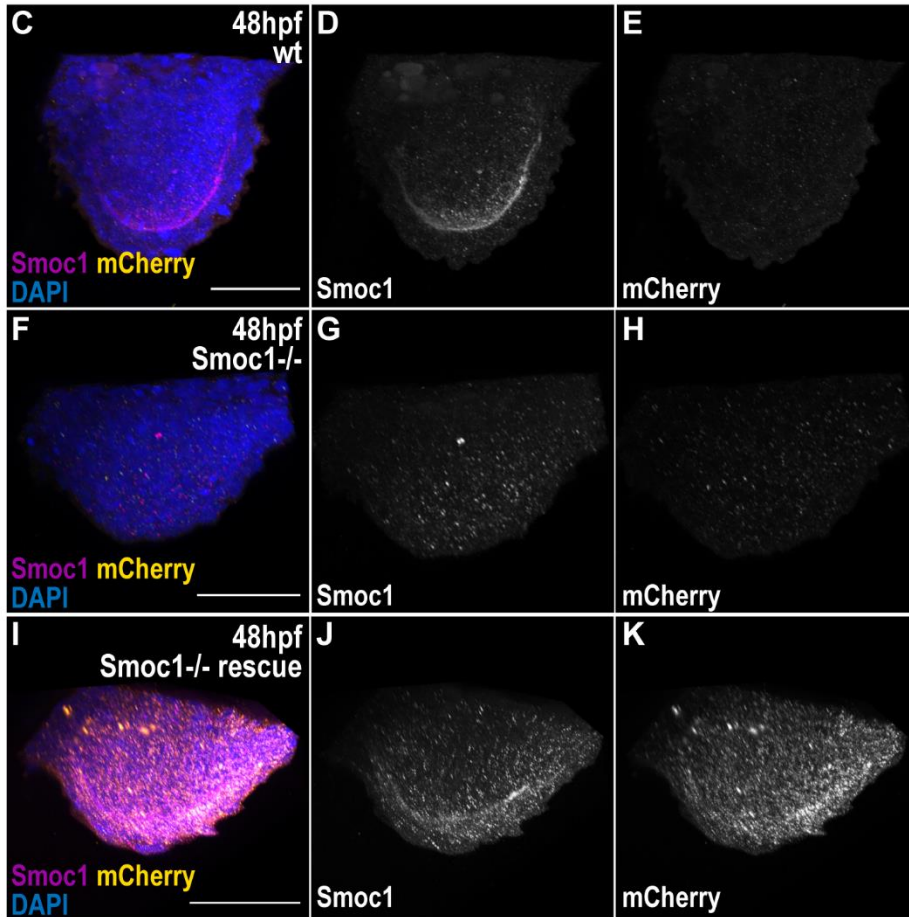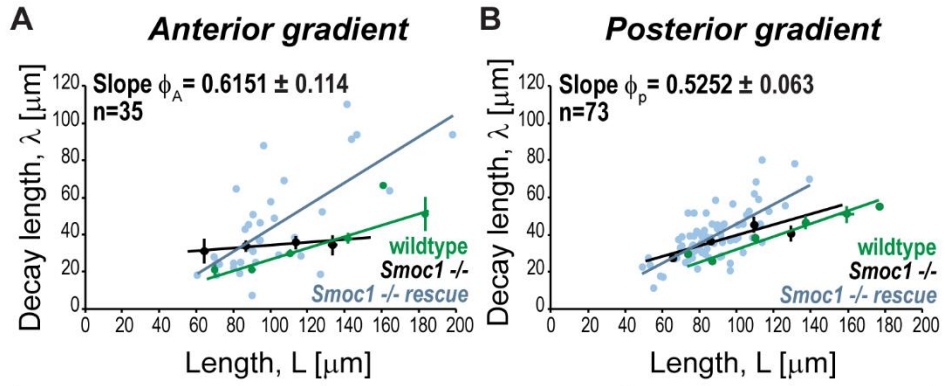
Figure S3. Smoc1 expression pattern in the pectoral fin. Related to Figure 4.
A-B, D-E, G-H Smoc1 immunostainings in pectoral fins at different times. Blue area, ROI. **C,F,I** Average intensity of normalized Smoc1 immunostaining signal *versus* relative position (along ROI midline). Light blue, individual profiles; dark blue, mean; shadowing, s.e.m. per relative position, Anterior, left; distal, down. **J-K** *smoc1* transcript pattern in pectoral fins (*in situ* hybridization using antisense probe) at 48 hpf. Dashed line, fin border. Asterisk highlights distal region of higher *smoc1* expression. **L** RT-PCR of cDNA of 24 hpf embryos in different conditions, including wildtype and two different concentrations of the splicing morpholino. Arrowheads, amplicons corresponding to the event of splicing block. **M** Sequences of wildtype, *Smoc1* heterozygotes and *Smoc1* homozygotes. Red box highlights nucleotide affected. **N** Western blots using mCherry (left) and Smoc1 (right) antibodies, performed in extracts from WT embryos and overexpression embryos injected with indicated mRNAs. Green asterisk highlights residual mCherry translation in embryos injected with *smoc1*^{ug104}-mCherry mRNA, which recapitulates the mutation of the CRISPR mutant. Because mCherry is C-terminal, this indicates a low level of read-through across the mutation caused by CRISPR, as previously reported (Schueren and Thoms, 2016). Green arrowheads indicate that the antibody against human Smoc1 recognizes the overexpressed zebrafish Smoc1. Anti-panActin antibody for the loading control. **O-T** Embryos highlighting the eye regions (dash) for different experimental conditions, at 30 hpf. **U** Average eye area for the conditions tested in O-T. For all statistical analyses: ***P<0.0001; two-tailed, unpaired, non-parametric Mann–Whitney tests. Mean ± s.e.m. are shown in all bar graphs. n represents number of fins analysed, except in U, where n equals number of eyes. Scale bars, 50µm.

A *Wildtype* Anterior gradient
$\overline{R}^2 = 0.67 \pm 0.04$
n=65

C Posterior gradient
$\overline{R}^2 = 0.75 \pm 0.03$
n=65

E *Smoc1 -/-* Anterior gradient
$\overline{R}^2 = 0.29 \pm 0.08$
n=98

G Posterior gradient
$\overline{R}^2 = 0.73 \pm 0.03$
n=98

I WT 60hpf PSmad1/5/9

J *Smoc1-/-* 60hpf PSmad1/5/9

K Anterior gradient PSmad1/5/9
Slope $\phi_A = 0.43 \pm 0.063$
n=50

L Posterior gradient PSmad1/5/9
Slope $\phi_P = 0.51 \pm 0.086$
n=34

M
Slope ϕ
*
wt -/-   wt -/-
Anterior   *Posterior*

N Anterior gradient PSmad1/5/9
Slope $\phi_A = 0.198 \pm 0.067$
n=36
wildtype
Smoc1-/-

O Posterior gradient PSmad1/5/9
Slope $\phi_P = 0.331 \pm 0.09$
n=19
wildtype
Smoc1-/-

P
Offset
**   *
wt -/-   wt -/-   wt -/-   wt -/-
Anterior *Posterior*   *Anterior* *Posterior*
BRE:GFP   PSmad1/5/9

Q 78hpf   MO control

R Anterior gradient MO control
Slope $\phi_A = 0.18 \pm 0.042$
n=21

S Posterior gradient MO control
Slope $\phi_P = 0.26 \pm 0.042$
n=34

T 78hpf   Smoc1 MO 2ng

U Anterior gradient Smoc1 MO
Slope $\phi_A = 0.12 \pm 0.174$
n=13

V Posterior gradient Smoc1 MO
Slope $\phi_P = 0.32 \pm 0.057$
n=26

W Anterior gradient *Smoc1-/-*
Slope $\phi_A = 0.077 \pm 0.086$
n=37

X Posterior gradient *Smoc1-/-*
Slope $\phi_P = 0.285 \pm 0.063$
n=63

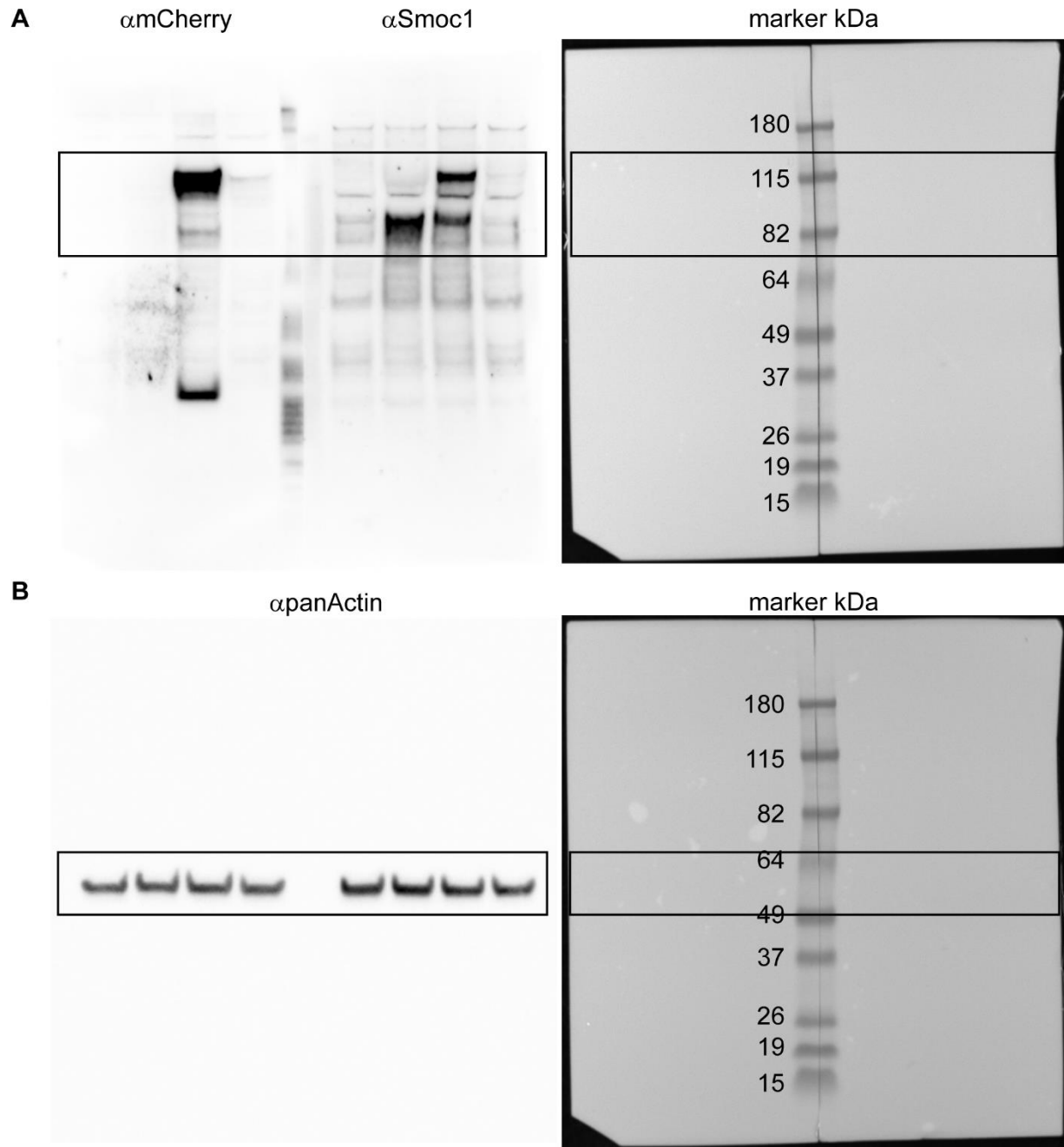**Figure S4.** *Smoc1^{ug104}* **mutant and morphants scaling phenotype. Related to Figures 3 and 5.**

A-H Gradient collapse analysis. Relative BRE:GFP concentration profiles $C(r,t)/C0(t)$ *versus* relative position in ROI, from 48 to 78 hpf, in wildtype (A, C) and *Smoc1-/-* mutants (E, G). Blue line represents the average curve of the profiles; standard deviation (shadows) per relative position is shown. **B, D, F, H** Density plots of the gradients in A, C, E, F. LUT corresponds to the fraction of normalized gradients passing through a certain bin. Bin size: 0.05 for ordinates and 0.05 for abscissas. **I-J** Phospho-Smad1/5/9 immunostainings showing gradients in *Smoc1* homozygote mutants (J) and wildtype controls (I) displayed as straightened ROI, like in Fig. 1D, at 60 hpf. **K-L** Decay length λ *versus* ROI half-length (L) of Phospho-Smad1/5/9 anterior (K) and posterior (L) wildtype gradients. Pink dots, individual data; black dots, average from length bins ± s.e.m. Lines, linear fits with goodness of fit: ϕ_A R^2=0.50; ϕ_P R^2=0.51. **M** Comparison of slope values, $\phi = \lambda/L$ from Phospho-Smad1/5/9 gradients in wildtype or *Smoc1-/-* conditions. **N-O** Decay length λ *versus* ROI half-length (L) in anterior (N) or posterior (O) Phospho-Smad1/5/9 gradients, in *Smoc1-/-* mutants (black and blue). Blue dots, individual data; black dots, average from length bins ± s.e.m. Black lines, linear fits with goodness of fit: ϕ_A R^2=0.22; ϕ_P R^2=0.48. Pink lines correspond to wildtype data from K-L. **P** Comparison of offsets from linear fits of scaling plots of BRE:GFP (from Fig. 3C-D, Fig. 5D-E) or Phospho-Smad1/5/9 (from K-L, N-O) gradients in wildtype and Smoc1-/- mutants. **Q-V** Gradient analysis in *smoc1* morphants. **Q,T** SEM images of pectoral fins of larvae injected with control morpholino (Q) or the *smoc1* splicing morpholino (T). Anterior on the left, distal down. **R-S, U-V** Decay length, λ, versus ROI half-length (L), in control morphants (R-S) or *smoc1* morphants (U-V). Lines, linear fits with goodness of fit: in controls ϕ_A R^2=0.53; ϕ_P R^2=0.56; *smoc1* morphants ϕ_A R^2=0.05; ϕ_P R^2=0.59. Note that the s.e.m. for the slope of the anterior *smoc1* morphant gradient is compatible with abolished scaling (slope = 0). **W-X** Decay length λ *versus* ROI half-length (L) in anterior (W) or posterior (X) BRE:GFP gradients of *Smoc1-/-* mutants (Figure 5C-D with individual data); Lines, linear fits with goodness of fit: ϕ_A R^2=0.02; ϕ_P R^2=0.26. In all scaling plots: slope (ϕ) values ± s.e.m. are shown. Mean ± s.e.m. are shown in all bar graphs. n represents number of fins analysed. For all statistical analyses: **P<0.01, *P<0.05; two-tailed, unpaired, non-parametric Mann–Whitney tests. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011). Scale bars: 20µm (I, J), 50µm (Q,T).

A — *Anterior gradient*

Slope ϕ_A = 0.6151 ± 0.114
n=35

Decay length, λ [µm]
Length, L [µm]

wildtype
Smoc1 -/-
Smoc1 -/- rescue

B — *Posterior gradient*

Slope ϕ_P = 0.5252 ± 0.063
n=73

Decay length, λ [µm]
Length, L [µm]

wildtype
Smoc1 -/-
Smoc1 -/- rescue

C 48hpf wt
Smoc1 mCherry
DAPI

D Smoc1

E mCherry

F 48hpf Smoc1-/-
Smoc1 mCherry
DAPI

G Smoc1

H mCherry

I 48hpf Smoc1-/- rescue
Smoc1 mCherry
DAPI

J Smoc1

K mCherry

L
Normalized Smoc1 intensity
Relative position
48hpf
wt n=7
Smoc1-/- rescue n=8

M
Normalized mCherry intensity
Relative position
48hpf
wt n=7
Smoc1-/- rescue n=8

**Figure S5. Smoc1[ug104] mutant rescue leads to gradient over-scaling. Related to Figure 5.**

A-B Decay length λ *versus* ROI half-length (L) in anterior (A) or posterior (B) gradients of *Smoc1-/-* mutants injected with 20pg *smoc1* mRNA (rescue; blue fit lines and dots). Slope (ϕ) values for rescue ± s.e.m. are shown. Lines, linear fits with goodness of fit: *Smoc1-/-* rescue ϕ_A R^2=0.49; ϕ_P R^2=0.51. Green corresponds to wildtype data from Fig. 3C-D; black represents *Smoc1-/-* data from Fig. 5D-E. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011). **C-K** Smoc1 (magenta) and mCherry (yellow) immunostaining expression in wildtype (C-E), *Smoc1-/-* mutants (F-H) or in *Smoc1-/-* mutants injected with 20pg of *smoc1-mCherry* mRNA (*rescue*) (I-K), at 48 hpf. Anterior on the left, distal down. **L-M** Average intensity of normalized Smoc1 (L) or mCherry (M) immunostainings *versus* relative position (ROI midline) in wildtype (blue), or Smoc1-/- mutants injected with 20pg *smoc1-mCherry* mRNA (rescue, yellow). Intensities were normalized to the wildtype maxima for comparison. Shadowing corresponds to s.e.m. per relative position. n represents number of fins analysed. Scale bars, 50µm.

Supplemental Data

A αmCherry αSmoc1 marker kDa

180
115
82
64
49
37
26
19
15

B αpanActin marker kDa

180
115
82
64
49
37
26
19
15

Data S1. Western blot source data. Related to STAR Methods and Fig. S3N.
Uncropped scans with size marker indications for Fig. S3N. **A** Uncropped scans of anti-mCherry and anti-Smoc1 blots. **B** Uncropped scan of loading control anti-panActin.
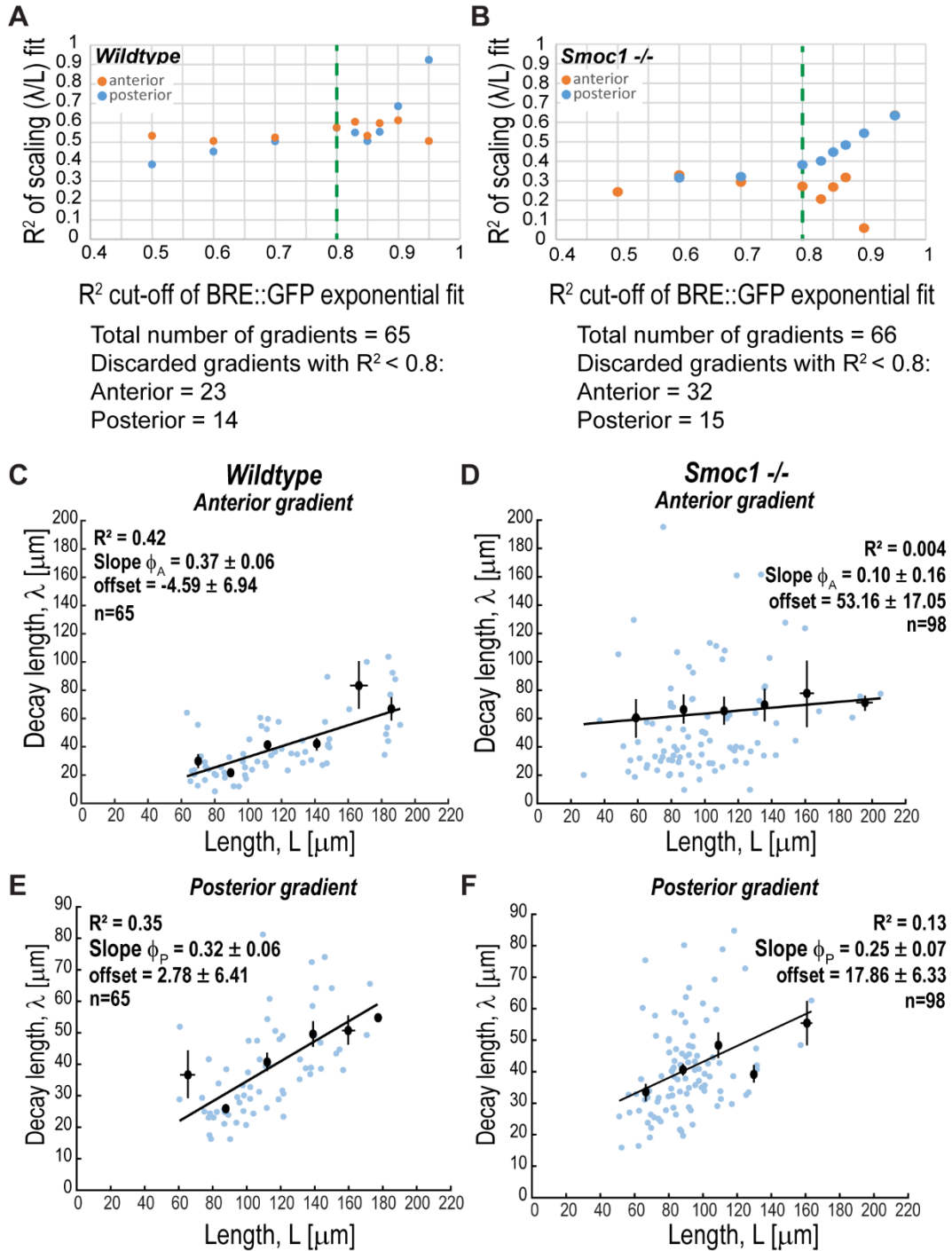
Data S2. Scaling of BMP signalling gradients (full datasets). Related to STAR Methods and to Figs. 3, 5, S4, S5.

A-B R^2 statistical analysis of BRE:GFP gradients. R^2 values obtained from linear fits in scaling plots (decay length, λ, *vs.* fin length, L) *versus* R^2 cut-off of individual BRE:GFP exponential fits, in wildtype (A) or *Smoc1-/-* mutants (B). Dashed lines, R^2 cut-off in which the quality of the scaling

plot fit improves considerably. Based on this, we considered for the scaling plot analysis only the profiles whose exponential gradients presented $R^2 \geq 0.8$. The estimation of decay length from fits with $R^2 < 0.8$ is likely to be inaccurate. **C-E** Decay length versus ROI half-length (L) in wildtype (C, E) and *Smoc1* homozygote mutants (D, F). Black lines, linear fits to individual data, with goodness of fit, R^2. Black dots, average from length bins ± s.e.m. Slope (ϕ) and offset values ± s.e.m. are shown. n represents number of fins analysed. BRE:GFP transgene used: BRE:eGFP (Laux et al., 2011).

Methods S1. Theoretical Framework. Related to Figs. 3, 5, S3, S4, S5.
1. Growth control and gradient scaling in two-dimensional geometries; 2. Amplitude power-law and growth control by relative time derivative of signaling; 3. Theory for ubiquitous expression of *smoc1.*

1. Growth control and gradient scaling in two dimensional geometries

I. HOMOGENEOUS AND ANISOTROPIC TISSUE GROWTH

We consider a tissue in three dimensions described by a coordinate system with the x-axis along the proximal-distal (PD) axis, the y-axis along the anterior-posterior (AP) axis and the z-axis along the dorsal-ventral (DV) axis. We discuss tissue growth that is is homogeneous and anisotropic and focus on the simple case where no growth occurs along the z-axis. The tissue size is characterized by the extensions along the main axies: L_x denotes the tissue size along the PD axis, L_y the size along the AP axis and L_z along the DV axis.

The area growth rate is

$$g = \frac{\dot{A}}{A} \tag{1}$$

The area growth rate can also be written as

$$g = \frac{\dot{L}_x}{L_x} + \frac{\dot{L}_y}{L_y} \tag{2}$$

where the dot denotes a time derivative. For homogeneous isotropic growth in the x-y plane, the cell velocity field is given by

$$v_x = \frac{g}{1+\epsilon}x \tag{3}$$

$$v_y = \frac{g\epsilon}{1+\epsilon}y \tag{4}$$

Here ϵ characterizes the growth anisotropy. For $\epsilon = 1$ growth is isotropic, for $\epsilon = 0$ growth is solely along the x-axis and in the limit of large ϵ growth is solely along the y-axis. The rates of change of tissue size then obey

$$\dot{L}_x = \frac{g}{1+\epsilon}L_x \tag{5}$$

$$\dot{L}_y = \frac{g\epsilon}{1+\epsilon}L_y \tag{6}$$

If ϵ is constant, the tissue sizes then scale as $L_x \sim L_y^{1/\epsilon}$.

II. GROWTH CONTROL BY RELATIVE RATES OF CONCENTRATION INCREASE

We have recently proposed a general principle of growth control based on the idea that increases in growth factor or morphogen levels stimulate growth at a rate that is proportional

to the relative rate of increase in concentration (Wartlick et al., 2011, 2014). A given cell uses receptors on its surface and a signaling pathway to detect changes in the concentration $c_{\text{cell}} = c(x_{\text{cell}}, y_{\text{cell}})$ at the position $(x, y) = (x_{\text{cell}}, y_{\text{cell}})$ of the cell. Here $c(x, y)$ is the morphogen concentration profile in space of a growth factor.

The growth at a rate g is generated as signaling output of the system with

$$g = \beta^{-1} \frac{\dot{c}_{\text{cell}}}{c_{\text{cell}}} \tag{7}$$

where

$$\dot{c}_{\text{cell}} = \partial_t c(x, y, t) + \mathbf{v} \cdot \nabla c(x, y, t) \tag{8}$$

is obtained using a time derivative in the reference frame that moves with the cells at velocity $\mathbf{v} = (v_x, v_y)$. It was shown before in a simpler case of one dimensional gradients that growth resulting from this growth control rule is spatially homogeneous if morphogen profiles scale (Wartlick et al., 2011, 2014).

III. SCALING OF THE MORPHOGEN PROFILE

We consider a morphogen profile that varies along the x axis only. Scaling of the morphogen profiles implies that if the tissue length L_x is rescaled by the tissue extension in x direction, the shape of the profile is not changed but the overall amplitude can be time-dependent. Scaling of the morphogen profile implies that it can be written as

$$c(x, y, t) = C_0(t) \, \Xi \left(\frac{x}{L_x(t)} \right) \tag{9}$$

Here $C_0(t)$ is the time dependent amplitude and the shape function $\Xi(r_x)$ with $r_x = x/L_x$ is independent of time. If the profile c scales and growth is homogeneous, we have $\dot{c}_{\text{cell}} = \dot{C}_0$. In this case the growth rate becomes

$$g = \beta^{-1} \frac{\dot{C}_0}{C_0} \tag{10}$$

independent of position in the three dimensional tissue.

If the amplitude $C_0(t)$ increases with system size as a power law with exponent q, we have

$$C_0 \sim L_x^q \sim L_y^{q/\epsilon} \tag{11}$$

we then find

$$\frac{\dot{C_0}}{C_0} = q\frac{\dot{L_x}}{L_x} = \frac{q}{\epsilon}\frac{\dot{L_y}}{L_y} \tag{12}$$

Using Eq. (2) we then have

$$g = \frac{1+\epsilon}{q}\frac{\dot{C_0}}{C_0} \tag{13}$$

This is consistent with the growth rule if $\beta = q/(1+\epsilon)$. Thus the signature of homogeneous growth due to a scaling morphogen profile is that the amplitude $C_0(t)$ increases with system size with an exponent q.

A. Scaling in the pectoral fin of the zebrafish

In order to test whether the profiles $c(x, y, t)$ observed in the fin actually scale during growth, we consider concentration profiles along the center line of a curved region of interest (ROI). The centerline shape can be defined as a time independent scaling function $y = L_y f(x/L_x)$ which for given value of x provides the y coordinate of the ROI centerline. We then consider the linear profiles along the ROI centerline

$$c_x(x, t) = c(x, L_y f(x/L_x)) \tag{14}$$

$$c_y(y, t) = c(L_x f^{-1}(y/L_y), y) \tag{15}$$

where f^{-1} denotes the inverse of the function f. If the profiles $c(x, y, t)$ scale, then

$$c_x(x, t) = C_0(t)\xi(x/L_x) \tag{16}$$

$$c_y(y, t) = C_0(t)\xi(f^{-1}(y/L_y)) \tag{17}$$

and the normalized profiles $c_x/C_0 = \xi(r_x)$ and $c_y/C_0 = \xi(f^{-1}(r_y))$ collapse each on a single curve. Fitting exponential functions $c_x/C_0 \simeq \exp(-r_x/\bar{\lambda}_x)$ and $c_y/C_0 \simeq \exp(-r_y/\bar{\lambda}_y)$ with fit parameters $\bar{\lambda}_x$ and $\bar{\lambda}_y$ to these collapsed profiles provides the scaling of decay lengths in x and y directions with system sizes $\lambda_x \simeq \bar{\lambda}_x L_x$ and $\lambda_y \simeq \bar{\lambda}_y L_y$.

IV. GROWTH CONTROL AND GRADIENT SCALING IN TWO DIMENSIONS

We consider a morphogen that is secreted within a localized source, diffuses isotropically and is degraded

$$\partial_t c + v_x \partial_x c + v_y \partial_y c = D\partial_x^2 c + D\partial_y^2 c - (k+g)c + s(x) = 0 \tag{18}$$

Here k denotes the degradation rate and D is a diffusion coefficient. Here the source is a stripe along the y-axis with $s(x) = \nu\theta(w_x - x)\theta(w_x + x)$, where w_x is the half source width and θ denotes the Heavyside function. We consider growth regulated by relative changes of concentration perceived by a cell in the dynamic tissue

$$g = \frac{1}{\beta}\frac{\partial_t c + v_x\partial_x c + v_y\partial_y c}{c} \tag{19}$$

If growth is homogeneous, the velocity fields are

$$v_x = \frac{g}{1+\epsilon}x \tag{20}$$
$$v_y = \frac{g\epsilon}{1+\epsilon}y \tag{21}$$

The system sizes in x was and y directions grow as

$$\dot{L}_x = \frac{g}{1+\epsilon}L_x \tag{22}$$
$$\dot{L}_y = \frac{g\epsilon}{1+\epsilon}L_y \tag{23}$$

We are looking for solutions to the dynamic equations with homogeneous growth that scale with system size

$$c(x,y,t) = C_0(t)\xi(\frac{x}{L_x}) \tag{24}$$

We thus have

$$0 = \frac{D}{L_x^2}\partial_{r_x}^2\xi(r_x, r_y) - (k + g(1+\beta))\xi(r_x) + s(r_x)/C_0(t) \tag{25}$$

where $r_x = x/L_x$. We can also write this as

$$0 = \partial_{r_x}^2\xi - \frac{k_g L_x^2}{D}\xi + \frac{sL_x^2}{DC_0} \tag{26}$$

where $k_g = k + g(1+\beta)$. Scaling solutions exist for $C_0 \sim L_x^2$ and $k_g \sim L_x^{-2}$. We thus find the solutions

$$c(x,t) = \frac{\nu}{k + g(1+\beta)}\left(1 - \frac{\sinh(L_x/\lambda - w_x/\lambda))}{\sinh(L_x\lambda)}\right)\cosh(x/\lambda) \tag{27}$$

for $x < w_x$, and

$$c(x,t) = \frac{\nu}{k + g(1+\beta)}\frac{\sinh(w_x/\lambda))}{\sinh(L_x\lambda)}\cosh((L_x - x)/\lambda) \tag{28}$$

for $x > w$, where

$$\lambda = \left(\frac{D}{k + g(1+\beta)}\right)^{1/2} \quad . \tag{29}$$

We thus have $C_0 \sim \nu/(k + g(1 + \beta))$. The growth law implies $\beta g = d \log C_0 / dt$. Since we also have $g = (1 + \epsilon) d \log L_x / dt$ and $C_0 \sim L_x^2$, scaling solutions exist for a specific value of $\beta = \beta_c$ with $\beta_c = 2/(1 + \epsilon)$.

2. Amplitude power-law and growth control by relative time derivative of signalling

Below is a summary of some theoretical considerations discussed in previous reports (Aguilar-Hidalgo et al., 2018; Wartlick et al., 2011, 2014), which are relevant to this work and presented here for convenience. We find an empirical power-law relationship between the gradient amplitude and the length of the organ as $C_0 \sim L_x{}^q$, with L_x the tissue length along the x axis. Because of scaling, the concentration of the morphogen in any cell in the tissue remains proportional to the amplitude, $C_0 \sim C_{cell}$ and therefore also $C_{cell} \sim L_x{}^q$. The time derivative of this expression gives $g_x = \frac{1}{q}\frac{\dot{C}_{cell}}{C_{cell}}$, with g_x the linear growth rate along the x axis, i.e. $g_x = \frac{\dot{L}_x}{L_x}$; and $\dot{C}_{cell}$ the time derivative of morphogen concentration in a cell. Considering a constant growth anisotropy $\varepsilon = g_y/g_x$, then the area growth rate is $g = g_x + g_y = (1 + \varepsilon)g_x$. Therefore $g = \frac{(1+\varepsilon)}{q}\frac{\dot{C}_{cell}}{C_{cell}}$.

To calculate the relative increase in cellular morphogen concentration $\alpha = \Delta C_{cell}/C_{cell}$ during a cell cycle of duration θ, we approximate $\dot{C}_{cell} \approx \Delta C_{cell}/\theta$ and $\theta \approx log(2)/g$. These approximations are fair, since $\dot{C}_{cell}$ and g change in time scales which are longer than θ. We then plug these approximations in the g equation above to obtain $\frac{log(2)}{\theta} = \frac{1+\varepsilon}{q}\frac{\Delta C_{cell}/\theta}{C_{cell}}$, from where $\alpha = \Delta C_{cell}/C_{cell} \approx log(2)\frac{q}{(1+\varepsilon)}$. In the fin bud, with $q = 0.9$ and $\varepsilon = 0.65$, then $\alpha = 0.38$. Therefore, fin cells divide when their BMP signaling levels increase by 38% from the beginning of the cell cycle.

3. Theory for ubiquitous expression of *smoc1*

In the main text, we showed that in Smoc1-/- mutants injected with *smoc1* mRNA, the gradient length scale expands as the tissue grows and appears to scale. The scaling factor in this experiment was larger than the one in control fins (Fig. 5F and Fig. S5A-B). In

this section we develop a theoretical framework within which to interpret these observations. Our logic is based on the assumption that Smoc1 acts as an expander of the morphogen by repressing its degradation as proposed in the expander-repression model by (Ben-Zvi, D. and Barkai, 2010). In the original expander-repression model, the expander forms an approximately flat profile and accumulates in the tissue over time because its production is repressed by the morphogen, it diffuses very fast and its degradation is negligible. Here, we show that this mRNA overexpression experiment recapitulates the circumstances of an expander-repression circuit.

Let us consider an expander molecule, E, that is transcribed everywhere in a tissue. The dynamics of E are then described by,

$$D_t E = \nabla(D_E \nabla E) - (k_E + g)E + s_E \tag{30}$$

Where D_E is the diffusion coefficient of E, k_E is the degradation rate of E, g is the tissue growth rate and s_E is the source term that captures the production rate of E. For the purposes of our analysis we assume that the diffusion coefficient is very large so that the expander is uniform in space. Under these conditions Eq. (30) can be rewritten as,

$$\partial_t E_0(t) = \frac{1}{L} \int_0^L \left[-(k_E + g)E_0(t) + s_E \right] \, dx \tag{31}$$

We can evaluate the integral on the RHS to obtain,

$$\partial_t E_0(t) = -(k_E + g)E_0(t) + \frac{w_E \, \nu_E}{L} \tag{32}$$

Where $E_0(t) = E(0, t)$. We can integrate Eq. (32) and use $E_0(t) = 0$ to obtain the following expression for the time dependent concentration of $E_0(t)$,

$$E_0(t) = \frac{w_E \nu_E}{L(k_E + g)} \left(1 - e^{-(k_E + g)t} \right) \tag{33}$$

For small $(k_E + g)$ we can use the approximation $e^{-(k_E + g)t} \approx 1 - (k_E + g)t$ which leads to,

$$E_0(t) = \frac{w_E \nu_E t}{L} \tag{34}$$

Following the original expander-repression model (Ben-Zvi, D. and Barkai, 2010), we assume that the expander degradation is negligible. At the same time, the growth rate in our system is very slow. It follows that Eq. (34) is approximately true. We are further assuming that

the expander is produced everywhere in the tissue (the mRNA is ubiquitously expressed) so $w_E = L$ and Eq. (34) reduces to,

$$E_0(t) = \nu_E t \tag{35}$$

It follows that the concentration of the expander accumulates over time. According to the expansion-repression model, the morphogen degradation rate, k , is regulated by the expander so that k decreases as the expander accumulates. We can therefore write the time dependent decay length of the morphogen as,

$$\lambda = \sqrt{\frac{D}{k(E_0(t))}} \tag{36}$$
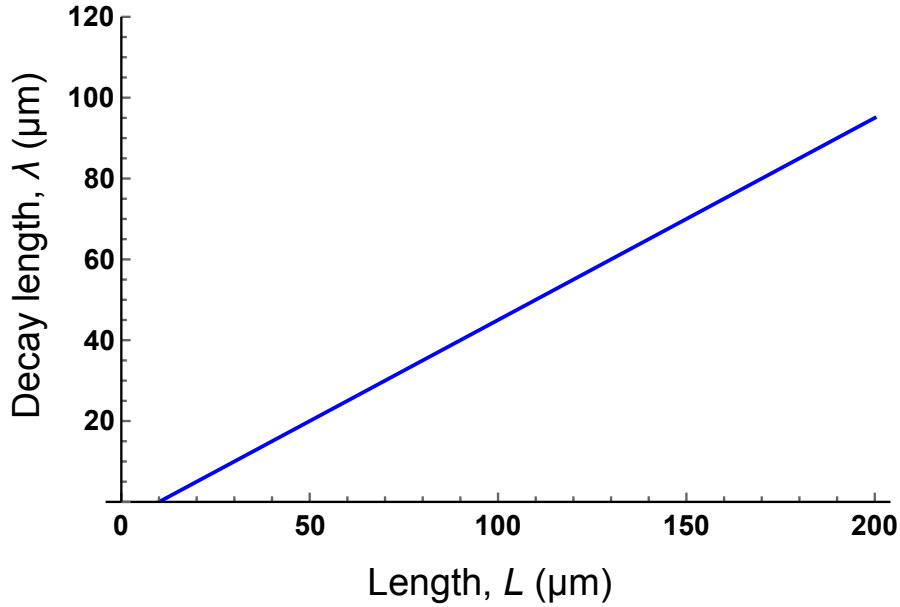
Methods Image 1. Theory of apparent scaling due to mRNA injection. The blue line shows the relationship $\lambda = \sqrt{\frac{D}{A}\frac{\nu_E}{v}} \left[L(t) - \hat{L} \right]$ for $\sqrt{\frac{D}{A}\frac{\nu_E}{v}} = 0.5$ and $\hat{L} = 10\mu m$.

At the same time, the embryo grows continuously during development so that its size is an increasing function of time, $L(t)$ (and therefore time is an increasing function of length, $t(L)$). It follows that λ becomes an increasing function of size which could in principle lead to the expansion of the gradient over time and approximate scale (as observed in the experiments). Methods Image 1 shows the predicted decay length versus size for a particular case where $k = \frac{A}{E_0(t)^2}$ and $L(t) = \hat{L} + vt$ for some constants A, $\hat{L}$ and v. In principle, qualitatively similar results are expected for other choices for the functions $k(E_0(t))$ and $L(t)$ as long as k

remains a decreasing function of the expander concentration and L is an increasing function of time.

In summary, we show that overexpressing *smoc1* by mRNA injection leads to the accumulation of the expander over time and could in principle scale the gradient. The dynamics of increasing expander concentration is in accordance with the expander-repression theory proposed by (Ben-Zvi, D. and Barkai, 2010). Much like in the expander-repression model, the mRNA injection leads to an increased concentration of the expander with the size of the tissue. This is expected to expand the gradient as the tissue grows since the expander suppresses the degradation of the morphogen.