HEART injury in adult mammals typically leads to permanent scarring. However, the adult zebrafish heart regenerates efficiently after injury, making zebrafish the preeminent model system for study of the cellular and molecular mechanisms of heart regeneration. After cryoinjury, which mimics aspects of myocardial infarction, the injured zebrafish heart undergoes a transient period of fibrosis, during which the damaged heart muscle regenerates via dedifferentiation and proliferation of cardiomyocytes. This process is accompanied by hypoxia-induced revascularization of the newly formed heart muscle tissue. Many pathways and factors involved in heart regeneration have been described. Important molecular signals for zebrafish heart regeneration emanate from the epicardium and endocardium. Other cell types are required for regeneration: inflammation is an early response to cryoinjury, and depletion of macrophages leads to delayed regeneration. Moreover, fibroblasts (potentially of different origins) are activated after injury, and ablation of fibroblasts leads to reduced cardiomyocyte proliferation. Altogether, these data suggest that cell types or transient activated cell states residing in the regenerative niche may be important cellular regulators of regeneration. A number of recent single-cell RNA sequencing (scRNA-seq) studies have determined cell type diversity in the mammalian heart, as well as in the regenerating zebrafish heart. However, to our knowledge, there has been no systematic data-driven attempt to identify proregenerative cell states and their cell type of origin. Consequently, our knowledge of the cellular composition of the regenerative niche and the underlying signalling interactions remains incomplete. Current definitions of activated macrophages and fibroblasts rely heavily on transgenes, which may be affected by observation bias and probably underestimate the complexity of cell states involved in regeneration.

Results
The cellular composition of the regenerating heart. For systematic identification of cardiac cell types in the healthy and regenerating zebrafish heart, we performed scRNA-seq of around 200,000 dissociated cells at different stages before and after injury. We observed a strong increase in fibroblasts and immune cells after injury. Closer inspection of the clustering data revealed a substructure among the cell types of the three main layers of the heart: epicardium, myocardium, and endocardium (Fig. 1a). We hypothesized that this cell type substructure might correspond to spatial differences due to the functional specifications of these cell types in the atrium and ventricle. Using the tomo-seq method for spatially resolved transcriptomics, we could validate atrial and ventricular enrichment for some of these cell subtypes (Fig. 1b). We confirmed this finding by physical separation of the atrium and ventricle, followed by scRNA-seq (Fig. 1c).
**Articles**

*d.p.i.* (days postinjury) (Fig. 2a) and partially colocalized with the hallmark of the regenerating heart, had increased in number by 3 established marker gene *nppa*

...with 13 transcriptionally diverse cardiac fibroblasts in more detail. Subclustering revealed mitogen which is the enzyme synthesizing retinoic acid; the cardiomyocyte

**Cell type diversity of cardiac fibroblasts.** We identified a further transcriptional substructure among the cardiomyocytes (Fig. 2a). In addition to adult cardiomyocytes, which are characterized by expression of genes involved in ATP synthesis and the tricarboxylic acid cycle (*atp5pd* and *aldoa*), we detected a smaller cluster characterized by genes associated with cardiomyocyte development (*tnn1*, *tnn2*, *bves* and *synpo2lb*), as well as *nppa*, which has previously been shown to be a marker for dedifferentiated cardiomyocytes in the border zone. These dedifferentiated cardiomyocytes, which are a hallmark of the regenerating heart, had increased in number by 3 d.p.i. (days postinjury) (Fig. 2a) and partially colocalized with the established marker gene *nppa* at 7 d.p.i. (Supplementary Fig. 3d).

We noticed that three well-established signalling factors in heart regeneration were strongly enriched in fibroblasts: *aldh1a2* (ref. 7), which is the enzyme synthesizing retinoic acid; the cardiomyocyte mitogen *mrgl* (ref. 30); and the proregenerative extracellular matrix (ECM) factor *fn1a* (Fig. 2b). This prompted us to investigate the diversity of cardiac fibroblasts in more detail. Subclustering revealed an unexpectedly large degree of diversity, with 13 transcriptionally distinct clusters of fibroblasts (Fig. 2c, Supplementary Fig. 4a,b and Supplementary Data File 3). These 13 clusters exhibited pronounced differences in their expression profiles of ECM-related genes (Fig. 2d), but their transcriptomic diversity extended far beyond that – the same fibroblast subtypes could be identified with high accuracy after removal of ECM-related genes (Supplementary Fig. 5).

**Identification of proregenerative cardiac fibroblasts.** To focus our analysis on those fibroblast subtypes that may be part of the regenerative niche, we analysed the dynamics of the cell clusters after injury. Three clusters of fibroblasts, characterized by expression of *col11a1*, *coll2a1* and *nppc*, respectively, were transiently present at the peak of regeneration (3 and 7 d.p.i.) but virtually absent before injury and after regeneration (Fig. 3a, Supplementary Fig. 4c and Supplementary Data File 4). Owing to their transient nature, we refer to these three clusters as cell states instead of cell types. Notably, *coll2a1*, a nonfibrillar collagen that may act as a matrix-bridging component, has already been shown to be expressed in epicardial and connective tissues following heart injury and is known to be

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**Fig. 1 | The cellular composition of the regenerating heart.**

*a* Cartoon of the experimental approach. Cells were barcoded with indels (insertions or deletions) during early development, and fish were raised to adulthood. Hearts were harvested either as an uninjured control or at 3, 7 or 30 d.p.i.

*b* UMAP representation of single-cell RNA-seq data and clustering results. Pie charts show the proportions of different cell types at different time points after injury. In the pie chart representation, similar cell types are grouped and shown by one (representative) colour. Asterisks denote cell types with a statistically significant change in proportions compared with uninjured controls.

*c* Mapping of single-cell data onto a spatially resolved tomo-seq data set. A computational deconvolution approach revealed chamber-specific cell subtypes.

*d* Distribution of subtypes of cardiomyocytes (CMs), endocardial cells and epicardial cells for scRNA-seq data sets in which atrium and ventricle were physically separated. Colour scheme as in Supplementary Fig. 3.
involved in regeneration of other organ systems in zebrafish\(^5\). Here, col12a1a was not only expressed in col12a1a fibroblasts but also in col11a1a fibroblasts, and these two transient fibroblast subtypes exhibited similar expression profiles overall (Supplementary Fig. 6).

Two other cell types with ECM-related functions, perivascular cells and valve fibroblasts, also showed increased abundance after injury (Fig. 3a, Supplementary Fig. 4c and Supplementary Data File 4). Other fibroblast types showed only moderate changes after injury. Importantly, established markers for ‘activated’ fibroblasts including postnb\(^3\) captured some but not all fibroblast clusters that are generated following injury and were also expressed in nonfibroblast populations including epicardial cells (Supplementary Fig. 6), suggesting that previous marker-based analyses have underestimated the transcriptional diversity of the cardiac fibroblast population.

To spatially resolve the identified fibroblasts, we performed fluorescence in situ hybridization (Fig. 3b) and Supplementary Figs. 7 and 8). This analysis confirmed the location of the transient fibroblast cell states in the border zone as well as the injury area. We next wanted to understand which of the transient fibroblasts expressed signalling factors involved in heart regeneration (Fig. 3c and Supplementary Fig. 9). We found particularly high expression of nrg1 in col12a1a fibroblasts, whereas fn1a was expressed almost exclusively in col11a1a fibroblasts. Furthermore, we observed that retinoic acid (aldh1a2) was produced at high levels by nppc fibroblasts as well as in the epicardium (Supplementary Fig. 8) and at lower levels in the endocardium (Supplementary Fig. 9). The retinoic acid readout gene stra6 (stimulated by retinoic acid 6)\(^4,5\) was expressed highly in col11a1a, col12a1a and proliferating fibroblasts (but not in cardiomyocytes), suggesting possible communication between these cell types. Moreover, we noticed a very specific interaction between perivascular cells and blood vessel endothelium via Cxcl12b–Cxcr4a chemokine signalling. Perivascular cells are known regulators of blood vessel formation\(^7\), and it was recently shown that Cxcl12b–Cxcr4a signalling is important for revascularization of the regenerating heart\(^7\).

We reasoned that a potential proregenerative function of transient cell states might be driven by secreted factors. Bioinformatic analysis revealed that expression of secretome genes increased in regenerating hearts at 3 d.p.i. and 7 d.p.i. compared with uninjured control hearts (Fig. 3d and Supplementary Fig. 10a). The col11a1a and col12a1a fibroblasts had the highest secretome expression of all detected cell clusters, and their secretome was enriched in genes with known functions in regeneration, morphogenesis and tissue development (Fig. 3e, Supplementary Fig. 10b and Supplementary Data File 5). The secretome at 3 d.p.i. was also enriched in genes with functions in angiogenesis (for example, angptl1a and angptl2a), suggesting a possible role of these cell types in revascularization. We next performed ligand–receptor analysis of the identified cell clusters (Methods). We found that the number of putative cell–cell interactions increased drastically after injury and peaked at 3 d.p.i., with a noticeable enrichment of both incoming (receptor) and outgoing (ligand) cell interactions in fibroblast subtypes (Supplementary Fig. 11).

Although our spatiotemporal analysis of single-cell gene expression profiles strongly suggested a proregenerative function of transient fibroblasts, additional experiments were needed to validate this hypothesis. To functionally assess the role of the transient col11a1a and col12a1a fibroblasts, we applied targeted genetic cell ablation using the nitroreductase/metronidazole (NTR/MTZ) system\(^7\). We generated and characterized a transgenic line

\(\text{Fig. 2 | Cell type diversity of cardiac fibroblasts. a. Left: relative changes in abundance of different subtypes of CMs across the time points (n = 3, 9, 9, 5 animals; error bars show s.e.m.). Right: differentially expressed genes between subtypes of CMs. b. Comparison of average normalized expression of known proregenerative factors in fibroblasts and other cell types. The comparison used data pooled over all time points, and fibroblasts were defined as all cells in the yellow cluster in Fig. 1b. c. UMAP representation of the subclustering of col11a1a-expressing cells. d. Expression of ECM-related genes in different fibroblast cell types. The genes were classified according to their contribution to structure, breakdown or interaction of the ECM. A, atrium; Ctrl., control; const., constitutive; dediff., dedifferentiated; prolif., proliferating; V, ventricle.}
Fig. 3 | Identification of proregenerative cardiac fibroblasts. a, Cell number dynamics of selected fibroblast subclusters across the time points (n = 3, 9, 5 animals; error bars show s.e.m.). b, Fluorescence in situ hybridization of marker genes. Left panel: const. fibroblasts (green), col12a1a fibroblasts (red) and nppc fibroblasts (purple) at 7 d.p.i. Right panel: col12a1a fibroblasts (red), col12a1a fibroblasts (white) and dedifferentiated CMs expressing ttn.2 (yellow) at 3 d.p.i. Injury areas (IA) are indicated with a dashed white line. Scale bar, 100 μm. c, Average expression of selected signalling genes in fibroblast subclusters. Blood vessel endothelial cells are included to show their interaction with perivascular cells via Cxcl12b–Cxcr4a signalling. d, Mean secretome expression of fibroblast cell types (n = 1,947, 8,611 and 17,261 cells; error bars indicate 3 × s.e.m.). e, Schematic of the ablation experiment for development or angiogenesis are highlighted (Bl. ves. EC: Blood vessel endothelial cells). f, Immunostaining of sections of cryoinjured hearts of Tg(-4kbcol12a1a:GAL4VP16;UAS:NTR:RFP) zebrafish treated with DMSO and MTZ at 7 d.p.i.; sections stained for Mel2c (CMs, red), PCNA (proliferation marker, green) and DNA (DAPI, blue). Arrowheads point to PCNA-positive cells. g, Labeled sections stained for Mef2c (CMs, red), PCNA (proliferation marker, green) and DNA (DAPI, blue). h, Percentage PCNA+ cells in control fish at 7 d.p.i. shows a significant decrease of PCNA+ cells in MTZ-treated fish (P = 0.0047). i, Histological comparison of the IA at 30 d.p.i. with and without MTZ treatment. j, Relative size of the IA across all histological replicates at 30 d.p.i. in 0.2% DMSO (n = 4) and MTZ-treated (n = 3) samples. n represents biologically independent samples from two independent experiments. Data are shown as mean and s.d. Two-tailed unpaired Student’s t test, P = 0.0027.

Labelling col12a1a-expressing fibroblasts, Tg(-4kbcol12a1a:GAL4VP16;UAS:NTR:RFP), subsequently referred to as col12a1a-NTR:RFP (Supplementary Fig. 12a–d and Methods). At 7 d.p.i., MTZ-treated col12a1a-NTR:RFP fish displayed a marked reduction in RFP expression, in line with our experimental design (Supplementary Fig. 12e). Whereas the injury area in col12a1a-depleted hearts at 7 d.p.i. looked normal overall (Supplementary Fig. 12f), five of six hearts displayed reductions in cardiomyocyte proliferation upon
**Fig. 4 | Identification of epicardial fibroblasts.** a. Cartoon of lineage tree construction using LINNAEuS. b. Weighted correlations of cell types over the tree were calculated to quantify lineage similarity. c, d. Clustering by lineage correlations at 3 d.p.i. (c) and 7 d.p.i. (d) revealed the epicardial origin of many niche fibroblasts. e. Cre-Lox lineage tracing confirmed the epicardial origin of col12a1a-expressing cells. Scale bar, 100 μm. f. Trajectory analysis suggested constitutive fibroblasts as the source of col11a1a and col12a1a fibroblasts at 3 d.p.i.
depletion of col12a1a-expressing cells compared with the vehicle control (Fig. 3g,h). At 30 d.p.i., we found that heart regeneration was significantly impaired by col12a1a+ cell ablation (Fig. 3i,j), suggesting that col12a1a-expressing fibroblasts contribute to regeneration. Although it was not possible to formally rule out the possibility that depletion of col12a1a-expressing cells may also lead to some tissue degradation of the ventricle independent of a regeneration defect, such effects would be expected to be minor, owing to the lack of col12a1a reporter expression in the uninjured ventricle (Supplementary Fig. 12c) and the lack of any visible degeneration outside the injury site (Supplementary Fig. 12e).

In summary, we identified three fibroblast states with potential proregenerative roles during heart regeneration: col11a1a, col12a1a and nppc fibroblasts. Our genetic cell ablation data strongly indicate a role for col12a1a-expressing cells in the regenerative niche. However, from which cell types these activated cell states are derived remains to be determined.

Identification of epicardial fibroblasts. We next aimed to elucidate the origins of the transient fibroblast states in order to better understand their mechanism of activation. Trajectory inference using partition-based graph abstraction (PAGA)⁴ suggested a dense network of possible transitions between all fibroblast subtypes (Supplementary Fig. 13). We reasoned that the large number of putative trajectories might be a computational artifact caused by a multitude of transcriptionally similar cell types, which might be generated by convergent differentiation pathways. We therefore decided to include explicit lineage information and analyse lineage relationships in a high-throughput manner using the LINNAEUS method⁵. In LINNAEUS, cells are marked by heritable DNA barcodes (genetic scars) that are created by Cas9 during early development (Methods) and whose distinct sequences identify cells that descended from the same parent at the time the scars were created. Here, we injected Cas9 into one-cell-stage embryos in order to record lineage relationships during early development until gastrulation⁶ (Fig. 1a), which we read out later in the adult heart (all lineage trees are shown in Supplementary Figs. 14–17). By sequencing scars and transcriptomes from the same single cells, we could build lineage trees that revealed the shared developmental origins of the cell types (Fig. 4a and Methods).

We first validated this approach in the adult heart using two control experiments. As expected, when preparing separate single-cell suspensions for the atrium and ventricle, the lineage tree showed separate atrial and ventricular branches, whereas a randomly split suspension did not show branch separation (Supplementary Fig. 18). We next sought to integrate the data across biological replicates in order to infer reproducible lineage relationships between cell types. In a lineage tree, all cells in a node share the same developmental ancestor, and any transient cell originating from a cell in a node can be found in the same node. We calculated correlations between cell type ratios in the different tree nodes to determine which cell types were related by lineage (Fig. 4b and Methods). Hierarchical clustering of these correlations revealed four clusters of cell types at 3 d.p.i. and seven clusters at 7 d.p.i. (Fig. 4c,d, Supplementary Fig. 19 and Methods). At both time points, all immune cells shared a common lineage, validating our approach. We observed a clustering of several fibroblast types, including col11a1a, col12a1a and constitutive fibroblasts, together with epicardial cells, strongly suggesting that these fibroblast cell types share a developmental origin with the epicardium (Fig. 4c,d).

To validate the lineage origins of col11a1a and col12a1a fibroblasts, we performed a genetic Cre–lox lineage tracing experiment in regenerating hearts using the transgenic line TgBAC(tcf21:Cre-ERT2; ubi:Switch). Our scRNA-seq data showed that tcf21 was expressed in epicardial cells, constitutive fibroblasts and other fibroblast types of the epicardial cluster (Supplementary Fig. 20). After recombination, epicardial cells, constitutive fibroblasts and other fibroblast types of the epicardial cluster. Furthermore, expression of tcf21 is not specific enough to address this question using Cre–lox lineage tracing (Supplementary Fig. 20), and reporter-based approaches reach their limits when analysing cell states with high resolution (our data did not identify suitable marker genes that could be used for lineage tracing of constitutive fibroblasts). To further elucidate the origins of the transient epicardial fibroblast states, we therefore applied a transcriptome-based trajectory inference method based on PAGA and RNA velocity⁷ to all ventricular cell types from epicardial lineage clusters at 3 d.p.i. and 7 d.p.i. (Fig. 4f, Supplementary Fig. 21 and Methods). At both time points, this approach showed that transient col11a1a and col12a1a fibroblasts originated from constitutive fibroblasts (Fig. 4f and Supplementary Fig. 21a) and suggested that the extremely transient col11a1a fibroblasts (which were only present at 3 d.p.i.) turned into col12a1a fibroblasts at 7 d.p.i. Genes that were upregulated along these trajectories were those encoding collagens col11a1a and col12a1a, proregenerative ECM factor fn1a, epicardial activation marker postnb, retinoic acid signalling response gene stra6 and cardiomyocyte mitogen nrg1 (Supplementary Fig. 21b,c).

Identification of endocardial fibroblasts. The transient nppc fibroblasts and several other fibroblast subtypes (spock3, cfd and valve fibroblasts) were not part of the epicardial lineage cluster at 3 d.p.i. or 7 d.p.i. but displayed moderate positive correlations with the endocardium and with each other. We reasoned that our correlation-based analysis had an underlying assumption that all clones would exhibit similar transition rates (Fig. 5a). This assumption may not be true for all cell types, as clones located in the uninjured part of the heart may not respond to the perturbation. When downsampling to 50% of the clones, we found that the cluster of epicardial fibroblasts remained stable, whereas the correlations among nppc, spock3, cfd and valve fibroblasts varied between iterations (Supplementary Fig. 22 and Methods). Such variability would...
Cell types that share a lineage

\[ \text{Clone-independent transition} \]

\[ \text{Clone-dependent transition} \]

Cell type overlap

Epicardium (V)
Fibroblasts (col12a1a)

Conditional cell type probability

\[ P(\text{Epicardium (V)}) = 14/14 = 1 \]

\[ P(\text{Fibroblasts (col12a1a)}) = 14/26 = 0.65 \]

Clone-independent transition

Fibroblasts (nppc)
Endocardium (frzb)
Endocardium (V)

Conditional cell type probability

\[ \text{< 0.8} \]

\[ 0.8 \]

\[ 0.9 \]

\[ 1.0 \]

Clone-dependent transition

Fibroblasts (col11a1a)

Figure captions:

**Figure a:** Lineage trees showing the relationship between different cell types. The trees illustrate both clone-independent and clone-dependent transitions.

**Figure b:** The overlap of cell types is depicted with 26 leaf nodes and 14 leaf nodes sharing specific lineage nodes.

**Figure c:** An example of cell type overlap with Epicardium (V) and Fibroblasts (col12a1a).

**Figure d:** Conditional cell type probability for Epicardium (V) and Fibroblasts (col12a1a).

**Figure e:** Gene expression and experimental setup, showing the use of Cre-ERT2, hsp70l, loxP, and 4-OHT for dissection and imaging.

**Figure f:** Dissection and imaging of cells, highlighting the expression patterns of Endocardium (V), Endocardium (frzb), and Fibroblasts (nppc).

**Figure g:** Upregulated genes comparing Fibroblasts (nppc) (7 d.p.i.) versus Endocardium (V) (7 d.p.i.) and Endocardium (V) (7 d.p.i.) versus Endocardium (frzb) (7 d.p.i.).
be expected if there were considerable differences between the activation profiles of the individual clones that make up these cell types.

We therefore developed a second computational approach, in which we analysed the conditional probabilities of co-occurrence of a cell type of interest with all other cell types (Fig. 5b and Methods). We reasoned that such a targeted approach would yield more reliable results specifically for cell populations of variable clonal behaviour. As validation, we found that col11a1a fibroblasts, constitutive fibroblasts and epicardium (ventricle) were present in >90% of the 3 d.p.i. lineage tree nodes that contained col12a1a fibroblasts (Fig. 5c). At 7 d.p.i., we found that the different endocardial cell types (endocardium (atrium), endocardium (ventricle) and endocardium (frzb)) were present in more than 80% of all nodes containing transient nppc fibroblasts, suggesting a lineage relationship between this fibroblast type and the endocardium (Fig. 5d). Similarly, the analysis suggested that spock3 and valve fibroblasts were also of endocardial origin, whereas cfd fibroblasts were of epicardial origin (Supplementary Fig. 23). Hence, our analysis revealed the existence of three endocardial-derived fibroblast subtypes (nppc, spock3 and valve fibroblasts), whereas all other fibroblasts shared a lineage origin with the epicardium. To validate our computational approach, we performed genetic lineage tracing using the transgenic line Tg(fli1a-Cre;ERT2; hsp70:Switch), which, after embryonic recombination, marks endothelial cells in the adult heart4,6. Expressed EGFP at 7 d.p.i. colocalized with endogenously expressed nppc (Fig. 5e), corroborating the endocardial origin of the transient nppc fibroblasts.

Only ~20% of the nodes with endocardium (ventricle) contained nppc fibroblasts (Supplementary Fig. 24a), suggesting that endocardial fibroblasts are only generated in a subset of the endocardium, potentially in the injury area. This clone-specific injury reaction of the endocardium would explain why targeted lineage analysis based on conditional probabilities outperforms correlation-based lineage analysis. We hypothesized that the internal position of the endocardium, as well as the transient nature and low clonality of the nppc fibroblasts, might mean that these cells are only generated from damaged endocardial clones. We confirmed that longer contact of the cryoprobe with the heart led to a larger injury area and resulted in much stronger nppc expression beyond the border zone, compared with injuries resulting from shorter contact times (Supplementary Fig. 24b and Methods).

Trajectory analysis revealed transcriptional similarity and a potential transition path between ventricular endocardium and nppc fibroblasts (Fig. 5f). We observed that nppc fibroblasts continued to express endothelial genes (for example vwf and fli1a) in addition to ECM genes, suggesting that they maintain at least part of their endocardial gene expression after turning on a fibrotic gene expression program. This prompted us to investigate in more detail what degree the gene expression profile of these cells had fibroblast characteristics. In particular, we considered whether the nppc fibroblasts might correspond to the activated endocardium that was previously described after injury13. We observed from our single-cell transcriptomic data that the endocardium underwent an activation response at 7 d.p.i. compared with healthy control hearts (Fig. 5g). However, this response was weaker than and only partially overlapping with the gene expression profile of nppc fibroblasts. Hence, the endocardium responds to injury with two separate transcriptional responses, which differ in the magnitude of their expression changes and which have not been systematically distinguished in the previous literature. Differentially expressed genes of nppc fibroblasts include many ECM-related factors (which led us to identify this cluster as a fibroblast state as shown in Fig. 2c) (Supplementary Data File 3). Furthermore, we noticed that nppc fibroblasts had upregulation of twist1b and express snail2, both of which are classical markers of epithelial-to-mesenchymal transition. This, together with the observation that some nppc cells appeared to be delaminating in our microscopy images (Fig. 5e), suggested that nppc fibroblasts may undergo at least a partial transition toward a mesenchymal state. However, as we did not observe a clear downregulation of VE-cadherin (cdh5) or upregulation of N-cadherin (cdh2), we speculate that nppc fibroblasts do not acquire full mesenchymal characteristics. Of note, it was previously reported that endocardial cells express collagen following injury but do not undergo a complete epithelial-to-mesenchymal transition14, whereas another publication showed that a minor proportion of endocardial cells appear to undergo a more complete transition15.

In summary, our analyses revealed a clear separation of epicardial- and endocardial-derived fibroblasts, with both lineages giving rise to distinct transient fibroblast states following injury. An endocardial origin for cardiac fibroblasts has also recently been established in zebrafish16,64, and our analysis provides strong evidence for the existence of fibroblasts of endocardial origin in the atrium and ventricle, as previously suggested in mouse62,64 and zebrafish14.

**Cellular dissection of the role of canonical Wnt signalling.** We noticed that fibroblasts expressed many genes related to Wnt signalling (ligands, receptors and modulators) (Fig. 6a and Supplementary Fig. 25), which inspired us to investigate the role of canonical Wnt signalling in this system. The role of Wnt signalling in heart regeneration remains an important open question65. On the one hand, Wnt is generally considered to be a proproliferative factor, and Wnt activation has been shown to be beneficial for zebrafish fin and spinal cord regeneration33,45. On the other hand, the role of Wnt signalling in cardiomyocyte dedifferentiation and proliferation is controversial46–48. We hypothesized that Wnt might exert its complex functions by regulating activation of prorenerative cell states, and we investigated whether cells with a shared lineage origin exhibited similar changes following a stimulus, that is, whether lineage relationship is predictive of perturbation response.

We therefore inhibited canonical Wnt signalling after cryoinjury using the well-characterized Wnt/β-catenin-dependent signalling inhibitor IWR-1 and observed the effects at 3, 7, 15 and 30 d.p.i. (Fig. 6b and Methods). Wnt/β-catenin signalling inhibition led to a significant delay in heart regeneration, with prolonged fibrosis and increased size of injury area compared with the control (Fig. 6b). Single-cell transcriptomics of IWR-1-treated hearts at 3 d.p.i. and 7 d.p.i. revealed that cardiomyocyte dedifferentiation was delayed (Fig. 6c). Compared with control samples, numbers of dedifferentiated cardiomyocytes were reduced at 3 d.p.i., and they did not localize to the injury area at 7 d.p.i. (Fig. 6c).

Levels of perivascular cells and all endocardial fibroblasts (nppc, spock3 and valve fibroblasts) were strongly reduced after Wnt inhibition, whereas the other transiently increased fibroblasts remained at similar levels overall (Fig. 6d, Supplementary Fig. 26 and Supplementary Data File 4). Specifically, the overall abundance of coll11a1a and coll12a1a fibroblasts did not change (Supplementary Fig. 26), even though coll12a1a fibroblast numbers peaked earlier after Wnt inhibition (at 3 d.p.i. instead of 7 d.p.i.); this may have been an indirect effect due to the observed impairment of heart regeneration. We confirmed these findings by fluorescence in situ hybridization (Fig. 6e). These observations strongly suggest that Wnt signalling is required for the activation of endocardial fibroblasts, akin to the described role of Wnt in inducing endothelial-to-mesenchymal transition in mice49,50.

We next wanted to understand how the absence of these cell types could mediate the observed delay in regeneration. Perivascular cells are known to exert a proangiogenic function through signalling and mechanical stabilization of nascent vasculature51. High expression of the chemokine cxc1l2b in perivascular cells in our data set (Fig. 3c) further supported the putative proangiogenic role of these cells. Furthermore, NPPC signalling has been shown to have a complex range of functions in the cardiovascular system52,53 and has...
therefore hypothesized that Wnt inhibition might delay heart regeneration via reduced revascularization following injury.

Recently been suggested to be involved in fin regeneration. We therefore hypothesized that Wnt inhibition might delay heart regeneration via reduced revascularization following injury. To assess revascularization, we quantified the proliferation of coronary endothelial cells (cECs) at 4 d.p.i. as well as the coverage of coronaries in the injury area at 7 d.p.i. (Methods). Following

| Table 1 | Relative Numbers of Cells Following Wnt Inhibition at 3 and 7 d.p.i. |
|---------|---------------------------------------------------------------|
| 3 d.p.i | 7 d.p.i |
| Ctrl    | 0.8 ± 0.2 | 0.6 ± 0.1 |
| IWR-1   | 0.6 ± 0.1 | 0.4 ± 0.05 |

Two-tailed unpaired Student’s t test, P = 0.0059.

**Fig. 6 | Cellular dissection of the role of canonical Wnt signalling.**

- **a.** Expression of Wnt signalling factors in different cell types of the zebrafish regenerating heart.
- **b.** Upper panel: cartoon summary of IWR-1 Wnt inhibition experiments. Lower left: histological comparison of the IA at 30 d.p.i. with and without Wnt inhibition. Scale bar, 300 μm. Lower right: relative size of the IA area as a percentage across all histological replicates; data shown as mean and s.d. Two-tailed unpaired Student’s t test between groups at each time point, P = 0.0059. Left: changes in relative numbers of dedifferentiated CMs at 3 and 7 d.p.i. between IWR-1- and DMSO-treated hearts (error bars indicate s.e.m.). Right: localization of dedifferentiated (tnn.2) cardiomyocytes at 7 d.p.i. with and without Wnt inhibition. Scale bar, 100 μm.
- **c.** Changes in relative numbers of non-CMs following Wnt inhibition at 3 and 7 d.p.i. (error bars indicate s.e.m.).
- **d.** Fluorescence in situ hybridization of perivascular cells (pdgfb, white) and nppc fibroblasts (nppc, purple) at 7 d.p.i. with and without Wnt inhibition. Scale bar in c and d, 100 μm. White dashed lines indicate the IA. **e.** Immunostaining of Tg(-0.8flt1:RFP) hearts at 4 d.p.i. with and without Wnt inhibition; RFP (coronaries, magenta), PCNA (proliferation marker, green) and DNA (DAPI, blue). Arrowheads point to PCNA (+) cECs in DMSO-injected (n = 5) and IWR-1-injected (n = 6) fish at 4 d.p.i. Data are shown as mean and s.d. Two-tailed unpaired Student’s t test, P = 0.012.
- **f.** Whole-mount images of Tg(-0.8flt1:RFP) hearts at 7 d.p.i. with and without Wnt inhibition. Orange dashed lines indicate injury area. Scale bar, 100 μm.
- **g.** Percentages of PCNA+ cECs in DMSO-injected (n = 5) and IWR-1-injected (n = 5) fish at 7 d.p.i. Data are shown as mean and s.d. Two-tailed unpaired Student’s t test, P = 0.02.

**Fig. 6 | Cellular dissection of the role of canonical Wnt signalling.**
administration of IWR-1, we observed a marginal but not significant reduction in cEC proliferation at 4 d.p.i. (Fig. 6f,g). However, at 7 d.p.i., we observed a clear and statistically significant reduction in the coverage of coronary vessels in the injury area (Fig. 6h,i). In summary, these results suggest that the observed reduction in heart regeneration following Wnt inhibition is at least partially mediated by a revascularization defect. This defect does not appear to be caused by reduced proliferation of CECs and might instead be linked to other aspects of revascularization, such as migration of coronaries or interactions with perivascular cells.

Discussion

Here, we used a combination of single-cell transcriptomics and spatiotemporal information to characterize putative proregenerative cell states in the adult zebrafish heart in a data-driven manner. This approach allowed us to identify three transient cell states with fibroblast characteristics (col11a1a, col12a1a and nppc fibroblasts) that are major sources of known proregenerative genes and also express additional secreted factors with potential proregenerative functions. These fibroblast cell states are ideally suited to be cellular drivers for regeneration based on their expression profiles, their location in the injury area and their transient appearance after injury. By combining high-throughput lineage tracing with RNA velocity and single-cell trajectory inference, we were able to systematically identify the origin of these cell states. This approach effectively combines information about two time points – CRISPR barcoding during early development and trajectory inference at the time of analysis. Furthermore, we developed two complementary computational approaches for integrating CRISPR barcoding data across biological replicates in order to obtain statistically sound information on lineage relationships between cell types. Taken together, the results obtained with these approaches can serve as a blueprint for understanding the origin of transient cell states that are generated under disease conditions.

Several lines of evidence (CRISPR lineage analysis, Cre–lox lineage tracing, transcriptome profiles and long versus short injury) showed that nppc fibroblasts were derived from the endocardium. The endocardium exhibited two separate activation programs that had not previously been systematically distinguished: nppc fibroblasts and a less pronounced endocardial activation profile. Functional experiments showed that activation of endocardial fibroblasts depends on canonical Wnt signalling. Wnt inhibition leads to a delay in regeneration which is accompanied by a revascularization defect. Although it is tempting to link the observed depletion of nppc fibroblasts and perivascular cells following Wnt inhibition to reduced revascularization, further experiments would be needed to substantiate this point. A combination of CRISPR lineage tracing, trajectory inference and Cre–lox lineage tracing showed that the col11a1a and col12a1a fibroblasts shared their lineage origin with the epicardium and were generated from constitutive fibroblasts following injury. Col11a1a and col12a1a fibroblasts have very similar transcriptional profiles, and our RNA velocity analysis suggested that col11a1a fibroblasts are precursors of col12a1a fibroblasts. Targeted cell type depletion of col12a1a-expressing cells after injury led to impaired heart regeneration, supporting the proregenerative function of these cells. In summary, our detailed cellular dissection of the fibrotic response in the zebrafish heart characterized sources and mechanisms of proregenerative programs. The large diversity of fibroblast cell populations found in the regenerating zebrafish heart suggests that fibrosis should be seen as a collective phenomenon to which a variety of cell types and states contribute.

The pathways that contribute to the regenerative capacity of the zebrafish heart are conserved in humans. Thus, our analysis may lead us to better understand the limited regenerative capacity of the human heart, while at the same time opening up an exciting strategy to identify potential therapeutic approaches. Our data set and the experimental and computational approaches presented here will serve as a powerful resource for identification of additional candidate interactions between cell states involved in heart regeneration.

Online content

Any methods, additional references, Nature Research reporting summaries, source data, extended data, supplementary information, acknowledgements, peer review information; details of author contributions and competing interests; and statements of data and code availability are available at https://doi.org/10.1038/s41588-022-01129-5.

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Single-cell RNA-seq and scar detection. Single cells were captured using a Chromium Single Cell 3′ kit (10X Genomics, PN-1000075), according to the manufacturer's recommendations. We aimed for 10,000 cells per library whenever possible. Both v2- and v3-chemistry were used for the data presented here. For simultaneous detection of transcriptome and CRISPR-induced genetic scar sequences for lineage tracing, we proceeded as previously described. In brief, the scar sequences were enriched by a two-round nested PCR approach, using 10 µl of the 10X Genomics CDNA and target site specific primers (dTomato transgene). The scar library was then indexed using the index primers of the Chromium kit to ensure they could be sequenced and demultiplexed together with the transcriptome. Samples were sequenced on an Illumina NextSeq 500 150 bp and an Illumina HiSeq 2500 200 bp after successful quality control by Bioanalyzer (DNA HS kit, Agilent).

Mapping and clustering of single-cell mRNA data. A zebrafish transcriptome was created with Cell Ranger 3.0 from GRCC11, release 92. Alignment and transcript counting of libraries were done using Cell Ranger 3.0. Library summaries are standardized in Supplementary Data File 1. The transcriptome data were filtered, clustered and visualized using Seurat 3.0 (ref. 31). We excluded the cluster of mpeg1.1 fibroblasts (Fig. 2c, i) from further analysis, as this could not rule out the possibility that the expression of the macrophage marker mpeg1.1 in this fibroblast cluster was an artifact caused by, for example, fragmentation of macrophages.

Histological staining, analysis and imaging. For the analysis of the injury area, animals were humanely killed at different times after injury by placing them in ice-cold water at 0–4 °C for 20 min. Hearts were dissected and incubated in 2 µl heparin and 0.1 M KCl in phosphate-buffered saline (PBS) for around 30 min. Cryosamples were fixed in 4% paraformaldehyde (PFA) in PBS overnight at 4 °C, washed in PBS for 3× 10 min and incubated overnight in 30% sucrose in PBS. Samples were then frozen in Tissue-Tek OCT Compound (Sakura) on a cryostat (Leica) to a thickness of ~200 µm. The cryo-sections were stained for myocyte and nonmyocyte markers (Thermo Fisher Scientific). For paraffin sections, tissue was fixed in 4% PFA in PBS overnight at 4 °C, washed in PBS for 3× 10 min and incubated in an ethanol series on 30% for 10 min, 20% for 20 min and 70% overnight at 4 °C. Next, samples were dehydrated in an ethanol series at 80% for 30 min, 90% for 30 min, 96% for 1 h, 100% for 5 min and 100% for 1 h, then in Toluid for 45 min and Toluid for 15 min, and finally embedded in paraffin. Connective tissue was stained using acid fuchsin orange G (AFOG). In brief, slides were dried for 30 min at room temperature. Next, slides were incubated in Bouin solution (Sigma-Aldrich) for 2 h at 60 °C and left for overnight incubation under the hood. Slides were washed for 30 min running double-distilled water (ddH2O) and incubated for 7 min in 1% phosphomolybdic acid (Sigma-Aldrich). Samples were washed for 3 min in running ddH2O and incubated with AFOG solution (made in our laboratory, Sigma-Aldrich) for 3 min. Slides were washed until clear with running ddH2O and rehydrated with 70%, 94% and 2× 100% ethanol and for 2× 5 min with xylol (Sigma-Aldrich). Slides were mounted with xylene mounting medium (Merck) and allowed to dry overnight under the hood. The analysis of injury size, the total ventricular tissue area and injury area on multiple slides were measured. Injury area was identified by visualization of collagen (blue) and fibrin (red). Images were obtained with a Keyence Microscope BZ-X800 using Viewer and analysed with ImageJ/Fiji (v.2.1.0).

RNAScope Multiplex Fluorescent v2 assay for immunofluorescence in situ hybridization and imaging. The RNAScope assay (ACD) for fluorescence in situ hybridization was used to localize different messenger RNAs (mRNAs) expressed by various cell types during heart regeneration. The technique was performed according to the manufacturer’s instructions for fixed tissues (323100-USM) or in combination with antibody staining (MK S1-150). Cryosections (7 µm) were dried for 30 min at room temperature before the experiment. The following changes were made: slides were incubated at 99 °C for 15 min with an antigen retrieval solution (ACD) using a steamer (WWM). Signal development was done using Tyramide Signal Amplification (TSA) plus fluorophores (PerkinElmer) fluorescein, cyanine 3 and cyanine 5 at 1:1,000 dilution. The probes used are listed in Supplementary Table 1. In addition, anti-EGFP (Sigma-Aldrich, catalog no. G1544, 1:500) and secondary antibody AlexaFluor 488 goat anti-rabbit IgG (Thermo Scientific, catalog no. A-11008, 1:500) were used in the protocol combining immunohistochemistry with in situ hybridization. The fluorescence in situ hybridization for the whole-mount embryo was performed according to the manufacturer’s instructions (MK 50-016). Images were obtained with a Zeiss LSM880 confocal microscope using Zen 2.3 SP1 FP3 (v.14.0.0.0) software and analysed using ImageJ/Fiji (v.2.1.0) and Photoshop (v.20.0.6) software.

Microscopy analysis of cardiomyocyte proliferation. Zebrafish hearts were extracted and fixed in 4% PFA in PBS overnight, washed in PBS for 3× 10 min and incubated overnight in 30% sucrose in PBS. Samples were then frozen in Tissue-Tek OCT Compound (Sakura) on dry ice. Tissues were cut at 7 µm on a cryostat (Leica) using SuperFrost slides (Thermo Fisher Scientific). For fluorescence, samples were washed with PBS (1×) before permeabilization with Antigen Unmasking Solution (Vector Laboratories, Inc. (H-3300), 1:100) at 80 °C for 20 min. Samples were then washed...
twice with PBST and incubated in blocking solution (5% BSA, 5% goat serum, 0.2% Triton X-100 in PBS), then incubated with primary antibodies overnight at 4°C, followed by two PBST washes and incubation with secondary antibodies for 2 h at room temperature. Slides were washed again with PBST and mounted with mounting medium Vectashield with DAPI (Vector Laboratories, Inc.). The primary antibodies used were anti-Mef2C (Santa Cruz Biotechnology, catalog no. sc-313, 1:200) and anti-PCNA (Dako, catalog no. M0879, 1:500). The secondary antibodies were AlexaFluor 488 goat anti-rabbit IgG (Thermo Scientific, catalog no. A-11008, 1:500) and goat anti-mouse IgG (Thermo Scientific, catalog no. A-21050, 1:500). Confocal images were taken using an LSM880 microscope (Zeiss).

The percentage of proliferating cardiomyocytes was calculated as a ratio of the number of (PCNA + Mef2C)2+ particles to the total number of Mef2C2+ particles in the injury area and in 200 µm of the injury border zone using ImageJ (v.2.1.0). Briefly, the total number of cardiomyocytes was analysed in the Mef2C2+ channel, as the total number of cardiomyocytes that did not count was using the article analysis tool after thresholding and applying a deepscaling filter. The number of (PCNA + Mef2C)2+ particles was determined using the particle analysis tool after merging the green and red (identified using look-up tables) channels and selecting a colour threshold for yellow. Thresholding was verified visually using the original image. For each biological replicate, three nonconsecutive midsectional sections were used.

Microscopy analysis of revascularization. Zebrafish ventricles were extracted from adults that had not been treated with morphine and fixed in 4% PFA for 1 h at 40°C. The ventricles were embedded in OCT and 8-µm sections were obtained (8 µm). For each biological replicate, three nonconsecutive midsectional sections were used. Immunostaining was performed as previously described24. The primary antibodies used were anti-RFP (Rockland, catalog no. 600-401-379, 1:200) and anti-PCNA (Santa Cruz Biotechnology, catalog no. sc-56, 1:500). The secondary antibodies were AlexaFluor 488 goat anti-mouse IgG (H+L) (Thermo Scientific, catalog no. A-11092, 1:500) and AlexaFluor 647 goat anti-rabbit IgG (H+L) (Thermo Scientific, catalog no. A-21244, 1:500). Confocal images were obtained using an LSM780 microscope (Zeiss). The percentage of proliferating cECs (PCNA2+) was calculated as the ratio of the total numbers of cECs in the injury area and in 200 µm of the injury border zone using ZEN 3.2 Blue software. For fluorescence intensity analysis, the percentage fluorescence was calculated from whole-mount images as a ratio to the background fluorescence in the injury area using Fiji software.

Wnt inhibition. To investigate the role of Wnt signalling in heart regeneration, the Wnt antagonist IWR-1 endo (Sigma-Aldrich) was used25. IWR-1 was dissolved in dimethyl sulfoxide (DMSO) to prepare a 10 mM stock solution. Wild-type fish were injected intraperitoneally with 25 µl of 10 µM IWR-1 in PBS or DMSO (0.1% in PBS). The injection was administered at 1 d.p.i. and 2 d.p.i. for the 3 d.p.i. analysis and, for later time points, once every 2 days from 2 d.p.i. until the day fish were euthanized.

Cloning and transgenic of 4-kb col12a1a:GAL4VP16. The possible 5′ upstream promoter region upstream of the predicted transcription start site of col12a1a was identified using Danio rerio strain Tuebingen chromosome 17, GRCz11 Primary Assembly as a reference sequence. Sequences were aligned using MUSCLE for primer design. Forward (5′-GAGAGAGGAGGAATG-3′) and reverse (5′-CTTGCTCACCATGATCGAAGCAG-3′) primers were designed using plasmid editor ApE and used to amplify the promoter region by PCR from genomic DNA of D. rerio. PCR was carried out in a 25 µl reaction using GoTaq Long PCR Mastermix (Promega, cat. no. M4021) according to the manufacturer’s protocol and TOPO cloned into pENTR5 (Invitrogen) to create pAM222. Tg(4-kbcol12a1a:GAL4VP16) is a multisite gateway assembly of pAM58, an empty clone. Expression was tested using GoTaq Green Mastermix (Promega, cat. no. G4029) and heat shocked for 1 h at 37°C. Adult fish were injured and hearts were harvested at 7 d.p.i.

Statistical and reproducibility data. Data collection and analysis were not performed blind to the conditions of the experiments. Sample collection and genomics analyses were performed by two independent experimenters. Sample sizes (n) are stated in each figure legend including the number of biologically independent samples over four independent experiments or in this section. Quantification of each experiment is described in the respective section of the Methods together with the software used. No data points were excluded from the analysis. Graphs and statistical analyses shown in Figs. 3b, i and 6b,i and in Supplementary Figs. 12f and 24b were performed in GraphPad Prism (v.7) and v.9) and the statistical tests in Figs. 3h,j and 6b,g,i and in Supplementary Figs. 12f and 24b were performed in GraphPad Prism (v.7 and v.9). Graphs show mean and s.d.; the statistical tests are indicated either in each figure legend including the number of independent samples used. The statistical tests in Figs. 3g used DMSO-treated fish (n = 3) and MTZ-treated fish (n = 6) at 7 d.p.i. n represents biologically independent samples from two independent experiments. The experiments in Fig. 3i used DMSO-treated (n = 4) and MTZ-treated (n = 3) fish; results are from two independent experiments. All RNAscope experiments in Fig. 3i, and c,e and in Supplementary Figs. 7, 8 and 12 were performed at least three times using one or two independent experiments, unless stated otherwise. n = 3, and 30 d.p.i., n = 8, and IWR-1 was injected at 3 d.p.i., n = 5; 7 d.p.i., n = 9; 15 d.p.i., n = 3, and 30 d.p.i., n = 8; n represents biologically independent samples over four independent experiments. In Fig. 6g, n represents biologically independent samples over two independent experiments; in Fig. 6i, n represents biologically independent samples over two independent experiments. The results in Supplementary Fig. 3d are based on n = 2 animals in one independent experiment, those in Supplementary Fig. 24c (n3 = 3, n4 = 3, n5 = 4, n6 = 3) are based on two independent experiments. The number of animals was determined based on previous literature4,16,39,40 for the experiments in Figs. 3b, 5e, 6f and 6i and in Supplementary Figs. 3d, 7, 8 and 12b,c or calculated using G*Power calculator1 by two-tailed t test with an error prob = 0.05 and power(1 - b err prob) = 0.80; effect sizes were determined as follows: Fig. 3g and Supplementary Fig. 12e,f = F2, Fig. 6b,c,e = F1.5; Supplementary Fig. 24c = f = 1.48. Mapping and filtering of single-cell scar data. Scar filtering was performed as described by Spanjaard et al.2. Briefly, scar sequences were aligned using bwa mem (version 0.7.17) to a reference genome (GRCz11) and variants were called using the single-cell transcriptome data (see previous paragraph). We removed reads that were unmapped, had an incorrect barcode or did not start with the exact PCR primer we used. We truncated all scar sequences to 75 nucleotides and removed that were unmapped, had an incorrect barcode or did not start with the exact PCR primer we used. We truncated all scar sequences to 75 nucleotides and removed that were unmapped, had an incorrect barcode or did not start with the exact PCR primer we used. We truncated all scar sequences to 75 nucleotides and removed that were unmapped, had an incorrect barcode or did not start with the exact PCR primer we used. We truncated all scar sequences to 75 nucleotides and removed that were unmapped, had an incorrect barcode or did not start with the exact PCR primer we used. We truncated all scar sequences to 75 nucleotides and removed that were unmapped, had an incorrect barcode or did not start with the exact PCR primer we used. This also reduced complexity for the subsequent filtering steps. In further filtering steps, we progressively filtered out scar sequences to remove all more than 15% mismatches that did not fulfill the following quality control conditions. As a first step, we required all molecules to be sequenced at least twice at remove straightforward sequencing errors; this also reduced complexity for the subsequent filtering steps.
To remove molecules caused by incorrect annealing, we selected only the most prevalent scar for each combination of barcode and unique molecular identifier (UMI), only the most prevalent UMI for each barcode and scar, and only the most prevalent barcode for each UMI and scar. We then assessed similar (Hamming distance of 2 or less) scar sequences in each cell, only keeping both if they met criteria designed to test whether both sequences were verifiably correct (see Spanjaard et al.4 for details). Finally, we removed cells that looked like doublets by comparing their numbers of different scars with the scar number distribution for that cell type.

Tree building. Tree building was done as described in Spanjaard et al.24. Briefly, the lineage tree building algorithm was designed to build lineage trees with high numbers of cells, despite the low detection rate inherent in single-cell sequencing. It consisted of two steps: we first determined the lineage tree in terms of scars created and afterwards placed the single cells in their appropriate positions in the tree.

To build the scar lineage, we used the data set of scars in single cells and determined which scars could be found in the same branch of the lineage trees: those scars were observed together in the same cell. We created a graph of scars with connections between scars, indicating that they were together in a branch. A scar that had been created early would be present in many branches and connected to many other scars. As a corollary, the scar that was connected to most other scars in this graph was created before all other scars. This scar became the root of the lineage tree; it was removed from the scar graph and we identified the most highly connected scar in the resulting graph. This scar was placed underneath the first scar in the lineage tree and was again removed from the graph. If at any point the graph split into disconnected parts after removing a scar, those parts were considered to be distinct branches of the lineage tree and were treated separately. This process was repeated until all scars had been placed in the lineage tree.

The above tree construction algorithm is valid for high, nonvariable scar-detection rates with perfect filtering and no doublets; for more realistic scenarios, computations are more complicated but follow the general scheme as outlined above. Refer to Spanjaard et al. for details4. Once the shape of the lineage tree and the location of all scars in the lineage tree had been determined, we placed the cells in the tree. Cells were placed using the scars that were found in them, at the lowest possible position in the tree. Importantly, this means that some cells may have been placed in leaf nodes, but typically some cells in which a leaf node scar had not been detected would be placed in internal nodes.

Lineage determination. To determine lineage relationships between cell types, we contracted all vertical branches (that is, removed all nodes that were not part of a lineage split) and then calculated cell type ratios — the amount of cells of a specific type divided by the node size — for each node. We calculated the lineage similarity between two cell types as the correlation between their cell type ratios, weighted by the size of the nodes. Importantly, this approach allowed us to integrate lineage data for multiple biological replicates. Using cell type ratios emphasizes nodes that are lower in the trees and contain fewer cell types, and weighting by node size emphasizes larger nodes to maximize reproducibility.

We determined the stability of clusters in two different ways. A thousand-fold repetition of downsampling all lineage trees to half of the cells was used to estimate the stability of lineage clusters against low cell numbers. After calculating the correlations and clustering the cell types, we recovered the same clusters in the majority of cases, indicating that these clusters were not influenced by low cell numbers (Supplementary Fig. 19). A thousand-fold repetition of downsampling all lineage trees to half of the nodes was used to estimate the stability of lineage clusters against low clonality or clone-specific transition rates. After calculating the correlations and clustering the cell types, we found that nppc, spock3, cfd and valve fibroblasts were attributed to varying clusters, indicating that differences between clusters were too large for the correlation analysis to yield reliable answers (Supplementary Fig. 22).

To determine possible sources of nppc, spock3, cfd and valve fibroblasts, we determined the overlap of cell types with these target cell types in tree leaf nodes. We calculated the conditional probability of cell types as the number of leaf nodes that contained cells of both source and target type divided by the number of leaf nodes that contained cells of the target type. For example, at 7 d.p.i., ventricular endocardium had a conditional probability of 1 with regard to nppc fibroblasts, meaning that at 7 d.p.i., every leaf node that contained nppc fibroblasts also contained ventricular endocardium. However, nppc fibroblasts had a conditional probability of 0.2 with regard to the ventricular endocardium, indicating an asymmetric lineage relationship between these cell types.

Trajectory analysis and RNA velocity. We used PAGA, implemented in scmap29 (v.1.8.2) and scvelo30 (v.0.2.4) for integrated trajectory and RNA velocity analysis. First, we extracted velocity information from reads aligned to a transcriptome based on GRCh11, release 92, using Cell Ranger 3.1 and velocyto30. Using cell type annotations determined previously, we selected relevant cells; regressed out total counts, numbers of genes and mitochondrial percentages; determined highly variable genes and principal components; and used knblk to integrate data sets. We then reclustered the data using the Leiden algorithm and ran PAGA on the Leiden clusters. For the integrated data set, we calculated RNA velocities using the stochastic model implemented in scvelo and plotted those on the UMAP representation of the data.

Secretome analysis. We performed an ortholog conversion of the Vertebrate Secretome Database2 using orthology data from the Alliance of Genome Resources, release 3.2.0 (ref. 73). For the ortholog conversion, we used all gene pairs with 1≤bestScore=Yes and kept all unique zebrafish gene names. After normalizing all cells to 10,000 transcripts, we calculated the percentage of transcripts that were in the transcriptome for each cell type at each time point. We then calculated log(1 + x) of the average expression of secretome genes per cell type and used the clustermap function in seaborn to perform hierarchical clustering using Ward’s method for the cell types and secretome genes that were expressed at an average of 10 in at least one cell type. We similarly performed hierarchical clustering of cell type averages of z transformed secretome expressions (capped at 5) for genes that had a z score greater than 2 in at least one cell type. We excluded neuronal cells and myelin cells from these analyses.

Interaction analysis. We used CellPhoneDB23 (v.2.1.4) to analyse interactions between cell types at each time point after ortholog conversion of the genes in CellPhoneDB, similar to the conversion of the secretome. We included all cell types with more than 50 cells or whose cell count proportion of the total data set exceeded 0.01 and counted the number of significant outgoing and incoming interactions for each cell type.

tomo-seq data deconvolution. We used AutoGeneS31 (v.1.0, https://github.com/theislab/AutoGeneS) to deconvolve the tomo-seq data. As input, we used the single-cell data constrained to the 5,000 most variable genes using scmap29 (pp.highly_variable_genes). AutoGeneS was then used to select a total of 400 informative genes from among the highly variable ones that differentiated the cell types. These highly variable genes were selected based on normalized dispersion. The proportions were inferred using nonnegative least squares32 (scipy.optimize.nnls) based on the cellular mean expression of the informative genes. For each bulk sample, negative proportions were set to zero and the rest were normalized to sum to one.

Reporting summary. Further information on research design is available in the Nature Research Reporting Summary linked to this article.

Data availability
Source data are provided with this paper. Sequencing data have been deposited in the Gene Expression Omnibus with accession numbers GSE159032 and GSE19619. Transcriptome data were aligned to a zebrafish transcriptome created with Cell Ranger 3.0.2 from GRCz11, release 92. We performed an ortholog conversion of the Vertebrate Secretome Database VerSeDa using orthology data from the Alliance of Genome Resources, release 3.2.0.

Code availability
Custom code is provided at https://github.com/Bastiaanspanjaard/Heart_regeneration.

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Author contributions

B.H., S.L., B.S., D.P. and J.P.J. conceived and designed the project. B.H. performed single-cell experiments and mRNA and cell type analysis, with support from J.M. and S.L. and M.G.S. did cryo-injury experiments. S.L. and A.M.M. generated and characterized the col12a1a>NTR:RFP line. S.L. performed and analysed histology. RNAseq, Cre–lox lineage tracing, pathway inhibition experiments and ablation experiments. B.S. performed secretome and cell–cell interaction analyses, built single-cell lineage trees, developed tree analysis methods, and performed trajectory and RNA velocity analyses. H.E. performed microscopy analysis of revascularization, with guidance from D.Y.R.S. H.A. adapted bulk deconvolution to tomo-seq data, with guidance from B.S. and F.T. D.P. guided experiments, and J.P.J. guided data analysis. B.H., B.S. and J.P.J. wrote the manuscript, in close interaction with S.L. and D.P. and with input from all other authors. All authors discussed and interpreted results.

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Competing interests

The authors declare no competing interests.

Additional information

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Software and code

Policy information about availability of computer code

- Data collection: Zen 2.3 SP1 FP3 (version 14.0.0.0) and Zen 3.2 blue edition were used to acquire confocal images. BIOREVO Keyence 82-9000 Viewer was used to acquire histological images.

- Data analysis: Cell Ranger (version 3.0.2) was used to align and demultiplex single-cell transcriptomics data. bwa mem (version 0.7.12) was used to map scRNA-seq data. Velocito (version 0.17) was used to extract mRNA velocity data from single-cell transcriptomics data. All data analysis was done in R (version 3.3 and higher), except when indicated. The R package Seurat (version 3.0) was used to cluster single-cell transcriptomics data. Interactions between cell types were calculated using CellPhoneDB (https://github.com/Techlab/cellphoneDB) (version 2.1.4) and further analyzed in custom scripts provided at https://github.com/Bastiaanspanjaard/Heart_regeneration. Analysis of cellular trajectories and mRNA velocity were done in Python (3.8) using scanpy (version 1.8.2) and scvelo (version 0.2.4). Deconvolution of TOMO-seq data was done in Python (3.8) using AutoGeneS (https://github.com/thesilab/AutoGeneS) (version 1.0). We developed custom code to filter, analyze and visualize scar data. The custom code is provided at https://github.com/Bastiaanspanjaard/Heart_regeneration. Microscopy data was analyzed in Fiji/magel including the Particle analyzer plug-in (version 2.1.0) and Adobe Photoshop (version 20.0.6). GraphPad Prism 7/9 (version 7.0d/9.3.1) was used for plotting graphs and statistical analysis.

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All manuscripts must include a data availability statement. This statement should provide the following information, where applicable:
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Sequencing data are deposited on Gene Expression Omnibus, accession numbers GSE159032 and GSE158919. Transcriptome data was aligned to a zebrafish transcriptome created with Cell Ranger 3.0.2 from GRCz11, release 92. We performed an orthologue conversion of the Vertebrate Secretome Database VerSeDa, using orthology data from the Alliance of Genome Resources, release 3.2.0.

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Sample size
The number of cells to be analyzed was based on the maximum number attainable. Since the various detected cell types have vastly different abundances, we focused the analysis on those cell types that were detected at high enough numbers for statistical analysis. Sample size for animal test groups was determined using G*Power calculator (G*Power 3.1).

Data exclusions
No data was excluded. However, the following filtering steps were applied to the scar sequencing data in order to mitigate the effect of sequencing errors, PCR amplification errors and formation of chimeric reads: removal of all sequences that were detected only once; removal of molecules with the same BC+UMI but different scar, same BC+scar but different UMI, and same UMI+scar but different BC; removal of scars with a Hamming distance ≤ 2 to a more highly detected scar; removal of the most lowly detected scars; and removal of cells that express more scars than expected (putative doublets).

For lineage analysis, only scars with creation probability <0.01 (as determined by bulk experiments) were used in order to minimize artifacts caused by repeated creation of the same scar sequence in one animal.

Replication
All attempts at replication were successful. The number of replicate experiments is provided in figure legends and/or the section “Statistics and reproducibility” in the Methods.

Randomization
Formal randomization of samples was not applicable, since no manual selection of single cells was performed during experiments and analysis. Animals were randomly assigned to test groups.

Blinding
Formal blinding was not applicable, since no manual selection of single cells was performed during experiments and analysis. Data collection and analysis were not performed blind to the conditions of the experiments. Sample collection and genomics analysis were performed by two independent experimenters.

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| Category                 | Description                                                                                                |
|--------------------------|------------------------------------------------------------------------------------------------------------|
| Laboratory animals       | Zebrafish (Danio rerio, strain AB). Adult zebrafish of random sex, aged between 4 months and a year, and with a length of at least 3 cm were used. |
| Wild animals             | No wild animals were used in this study.                                                                   |
| Field-collected samples  | No field collected samples were used in this study.                                                        |
| Ethics oversight         | Zebrafish were bred, raised and maintained in accordance with the FELASA guidelines, the guidelines of the Max-Delbrück Center for Molecular Medicine and the Max Planck Society, and the local authorities for animal protection (Landesamt für Gesundheit und Soziales, Berlin, and The Veterinary department of the Regional Board of Darmstadt, Darmstadt, Germany) for the use of laboratory animals based on the current version of German law on the protection of Animals and EU directive 2010/63/EU on the protection of animals used for scientific purposes. In addition, housing and breeding standards followed also the international ‘Principles of Laboratory Animal Care’ (NIH publication no. 86-23, revised 1985). |

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