A multiomic atlas of human early skeletal development

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Abstract

Bone and joint formation in the developing skeleton rely on co-ordinated differentiation of progenitors in the nascent developing limbs and joints. The cell states, epigenetic processes and key regulatory factors underlying their lineage commitment to osteogenic and other mesenchymal populations during ossification and joint formation remain poorly understood and are largely unexplored in human studies. Here, we apply paired single-nuclei transcriptional and epigenetic profiling of 336,000 droplets, in addition to spatial transcriptomics, to construct a comprehensive atlas of human bone, cartilage and joint development in the shoulder, hip, knee and cranium from 5 to 11 post-conception weeks. Spatial mapping of cell clusters to our highly multiplexed in situ sequencing (ISS) data using our newly developed tool ISS-Patcher revealed new cellular mechanisms of zonation during bone and joint formation. Combined modelling of chromatin accessibility and RNA expression allowed the identification of the transcriptional and epigenetic regulatory landscapes that drive differentiation of mesenchymal lineages including osteogenic and chondrogenic lineages, and novel chondrocyte cell states. In particular, we define regionally distinct limb and cranial osteoprogenitor populations and trajectories across the fetal skeleton and characterise differential regulatory networks that govern intramembranous and endochondral ossification. Through somatic mutation analysis, we predict two new potential cell origins for human chondrocyte development. We also introduce SNP2Cell, a tool to link cell-type specific regulatory networks to numerous polygenic traits such as osteoarthritis. We also conduct *in silico* perturbations of genes that cause monogenic craniosynostosis and implicate potential pathogenic cell states and disease mechanisms involved. This work forms a detailed and dynamic regulatory atlas of human fetal skeletal maturation and advances our fundamental understanding of cell fate determination in human skeletal development.

Main

Human bone formation begins between 6-8 post-conception weeks (PCW) in the period spanning the transition from embryonic to fetal stages. Within the calvaria of the cranial skeleton, progenitors differentiate into osteoblasts through intramembranous (IM) ossification and expand from ossification centres that eventually meet to establish the prenatal suture joints that house osteoprogenitors [1,2]. In the appendicular skeleton, the nascent synovial joint first appears as an interzone condensation in the limb bud at 5-6 PCW [3], which subsequently forms a cavitated region that articulates adjacent incipient cartilage templates. The latter acts

as a temporary scaffold to facilitate development of the body plane and is gradually replaced

by bone tissue as development proceeds through endochondral (EC) ossification [4,5].

These two distinct and anatomically restricted modes of ossification govern osteogenesis and

joint formation throughout the human skeleton and, to our knowledge, the cellular bases by

which they form and mature remain partially described in human development at single-cell

resolution. To address this, we applied single-nuclei paired RNA and ATAC sequencing, and

multiple spatial methods to decipher the gene regulatory networks that mediate maturation of

the distinct bone and joint-forming niches in the cranium and appendicular skeleton across

space and time from 5-11 PCW. Through this, we discovered previously undescribed cellular

diversity in the osteogenic and chondrogenic lineages. We develop ISS-Patcher, a tool to

impute cell labels from the droplet data on our high-resolution 155-plex in situ sequencing

(ISS) datasets, allowing us to gain detailed insights into spatially defined niches within the

embryonic synovial joint. Applying OrganAxis, a new spatial transcriptomics annotation tool,

we also define the spatial trajectory of the developing frontal bone of the fetal skull.

In addition, our comprehensive resource and new computational toolset including SNP2Cell

enabled us to gain new insights into the molecular mechanisms of developmental bone

disease, such as craniosynostosis[6–8] as well as to implicate region-specific contributions of

bone and cartilage lineages in ageing diseases of the human skeleton, such as

osteoarthritis[9,10].

Results

Cellular taxonomy across first trimester skeletal joint development

We applied paired droplet-based snRNA-seq and snATAC-seq (10x Genomics Multiome) to

define the cellular taxonomy of the developing embryonic fetal skeleton between 5-11 post

conception weeks (PCW) across 12 fetal donors. Focusing on three major synovial joints, we

sampled whole-intact nascent and maturing shoulder, hip and knee joints across the

developmental timepoints (Fig. 1a-b). For the developing skull and cranial suture joints, which

had not been profiled in human across embryonic stages, we studied the anterior and posterior

regions of the cranium separately, and sampled the calvaria and skull base individually to

divide intramembranous and endochondral bone-forming niches. (Fig. 1a and Supplementary

Table 1). We captured 336,162 high-quality transcriptome and chromatin-accessibility droplets

and curated the dataset from all regions into eight shared major cellular compartments based

on marker gene expression: mesenchymal, muscle, immune, endothelial, Schwann, neural,

epithelial, and erythroid cells (Fig. 1c-e, Extended Data Fig. 1c). High concordance was

96 observed between the transcriptome and ATAC peak compartment structures (Extended Data

97 Fig. 1f), and we observed that mesenchymal cells were the most abundant across all regions,

98 while myogenic cells were absent in the calvaria (Fig. 1e). From these we defined 30 broad

99 cell clusters and conducted sub-clustering of lineages which enabled the identification of 122

100 fine-grained clusters (Supplementary Table 2).

102 Previous single-cell (scRNA-seq) atlases of developing human and mouse limbs [11–13] profiled

103 whole cells and captured low numbers of maturing osteoblast transcriptomes and *COL10A1+*

104 hypertrophic chondrocyte transcriptomes. To facilitate the reconstruction of developmental

105 trajectories in osteochondral lineages, we profiled nuclei-droplets from a large number of cells.

106 This captured comparatively more diverse chondrogenic and osteogenic subcompartments,

107 suggesting that matrix-rich stromal populations are relatively resistant to enzymatic digestion

108 and less amenable to whole-cell profiling [11–13].

110 Our approach enabled the discovery of osteogenic cell state trajectories enriched in the

111 appendicular joints and skull base (formed through endochondral ossification) and calvarium

112 (formed through intramembranous ossification), reflecting different osteoblast origins from

113 anatomically distinct regions (Extended Data Fig. 2). Chondrogenic clusters were relatively

114 depleted in the calvarium (Fig. 2a), consistent with the mechanisms of intramembranous bone

115 formation[14]. In addition, we uncovered novel cell states of the facial and pharyngeal regions,

116 described *COL10A1*-expressing fetal hypertrophic chondrocyte development, discovered a

117 novel *PAX7+* chondrocyte population, and described a potential pathway leading to its

118 formation from common myogenic progenitors[15]. We delineate the previously unreported

119 process of human Schwann cell development and the formation of fetal *PI16+* fibroblasts, the

120 latter of which shares transcriptional similarities with postnatal pan-tissue perivascular

121 fibroblasts.

123 Sequencing-based spatial transcriptomics previously deployed on the embryonic limb has

124 highlighted challenges in capturing transcripts within the early ossifying bone and cartilaginous

125 precursors (cartilage anlagen)[11]. In order to resolve bone lineage cell states in space and

126 understand organisation of the nascent synovial joint that bridges the adjacent anlagen, we

127 performed high resolution 155-plex ISS of the whole intact early embryonic upper (6.7 PCW)

128 and lower (5.7 PCW) limbs, late embryonic (7.3 PCW) knee, and shoulder regions (Fig. 1f and

129 Extended Data Fig. 2a-c). Additionally, we conduct sequencing-based spatial transcriptomics

130 (10x Genomics Visium CytAssist) of the developing coronal suture (9 PCW) and frontal bone

131 (Extended Data Fig. 2d) allowing us to capture the trajectory of osteo-lineage development

132 across space. We leveraged these spatial data to systemically curate cell lineages within the

133 mesenchymal compartment in a well-defined spatial context.

Spatiotemporal zonation of the incipient synovial joints

Synovial joint-site determination occurs between 5-6 PCW in the limbs, orchestrated by the
emergence of a mesenchymal condensation comprising *GDF5*-expressing populations
termed the interzone (IZ). Part of them undergo differentiation into joint-articulating
chondrocytes at the ends of the endochondral bone primordia, and chondrocytes that adopt a
hypertrophic phenotype in the cartilage anlagen[16]. To identify cell-states that constitute these
early joint-specific progenitors, we applied differential abundance testing focusing on the non-
myogenic mesenchyme across developmental time (5-11 PCW) and skeletal regions
(shoulder, hip, knee) (Fig. 2a). We identified four progenitor populations enriched in the early
appendicular joint: Interzone Chondrocytes (InterzoneChon), *HIC1*+ mesenchyme (*HIC1*+
Mes), *PI16*-expressing fibroblast-progenitors (FibroPro) and dermal fibroblasts (DermFibro).
These clusters support the notion that numerous early progenitors exist between 5-7 PCW
and give rise to the chondro-fibro lineages in the joint, instead of a single master *GDF5*+
progenitor population[17].

We first focused on deciphering the composition of progenitors in the early synovial joint,
where fibrous ligaments[18], tendon[19] and cartilage components are thought to derive from a
progenitor *GDF5*+ population in mice[17,20]. To this end, we performed sub-clustering of the
broad InterzoneChon (*GDF5*+) population and leveraged RNA-velocity to infer their pseudo-
trajectory (Fig. 2b-c). We then applied SCENIC+ to identify predicted gene programs and TF
accessibility changes across the resulting subclusters (Fig. 2d). Our analysis revealed multiple
subclusters with specific tissue expression signatures, including an early *PRRX1*+
mesenchyme population (EarlyIZ) enriched for mesenchyme-associated signatures (Fig. 2b).
Early IZ was predicted to express TFs associated with early limb development (e.g. *TBX18*,
SHOX, *LHX9*)[11] and demonstrated low *RUNX2* expression and accessibility in downstream
target regions but moderate *SOX5, SOX6* and *SOX9* expression and target region
accessibility, suggesting a poised trajectory favouring chondrogenesis over osteogenesis (Fig.
2d).

We defined four main trajectories that emerge from this latter population, characterised by
transcriptional signatures of fibroblasts (Fibro IZ), articular (Articular IZ) and hypertrophic
chondrocytes (Hypertrophic IZ-1, Hypertrophic IZ-2), and flanking dermal mesenchyme
(Flanking IZ) (Fig. 2b-c). Three clusters; Articular, Fibro and *GDF5*high IZ, expressed *GDF5*.
Articular IZ was more prevalent in the knee joint, whereas Fibro IZ is enriched in the ball-and-
socket joints of the shoulder and hip (Fig. 2c). Articular IZ particularly expressed chondrogenic

markers (*COL2A1, ADGRG6, PIEZO2, LGR5, NOG*) comparable to mouse IZ progenitors for articular chondrocytes [17,20] including *ENPP1*, which negatively regulates chondrocyte hypertrophy and bone formation, in keeping with a role in articular cartilage formation. While *GDF5*-high IZ was predicted to highly express *CREB5* and *DBX2* and its target genes, the latter of which is associated with digital IZ formation regulated by *HOX* genes[17,21], chondrogenesis TFs were not highly expressed, and predicted TF activity and accessibility for *RUNX2* was lowest among the IZ populations. We therefore hypothesise that *GDF5high* IZ is maintained as a progenitor pool, potentially to sustain continuous influx into the forming joint[18]. In the *GDF5 negative* hypertrophic clusters, progressive TF expression and accessibility of *RUNX2*[22–24], and concurrent downregulation of articular chondrocyte TFs (*CREB5*, EGR1) signifies a hypertrophic phenotype (Fig. 2d). Hypertrophic2 IZ transitions through Hypertrophic1 in the inferred trajectory and the former expressed genes that are upregulated during formation of anlagen or pre-hypertrophic chondrocytes in mice (*BMP5, HAPLN1, ACAN* and *ZFHX3*)[19,25], suggesting they form the incipient cartilage template. Overall, our *in vivo* data are consistent with *in vitro* observations of the propensity for human-iPSC and mouse-ESC derived *GDF5*+ stromal cells to form an articular, rather than a hypertrophic phenotype [26,27].

To understand spatial zonation of the nascent synovial joint we next visualised the pre-cavitating knee joint ISS to identify regions of incipient hypertrophic anlagen (*RUNX2)*, early progenitors (*TBX15, SHOX)* and nascent articular cartilage and interzone (*SOX9*) (Fig. 2e). We then leveraged the snRNA-seq dataset and imputed genes missing in the 155-plex ISS dataset through our newly developed ISS-Patcher function (see Methods) to infer cell labels to the 155-plex clustered manifold (Fig. 2f,g). In the embryonic lower limb, Early IZ was diffusely distributed across regions of the interzone and anlagen, and was surrounded by Flanking IZ (Fig. 2c,d). Articular IZ (*GDF5*+), were predominantly enriched in sites of incipient knee articular cartilage formation which also showed SOX9 staining (Fig. 2e,g). In contrast, Fibro IZ (*GDF5*+) was enriched in the shoulder interzone region adjacent to the articular surface of the humerus (Fig. 2g), and expressed *SCX* (Fig. 2b) which plays a role in ligament formation in the IZ in mice[28]. The differential enrichment and signatures of the Articular and Fibro IZ populations may be explained by the need for fibro-cartilage labrum formation in the ball-and-socket joints of the shoulder and hip which emerge and mature later in development. We coalesce the trajectories and spatial enrichment of the newly defined interzone progenitors and describe a model for zonation of the embryonic joint. Within the model, the future joint region (Articular phenotype) forms in the centre of the interzone defined by early pro-chondrogenic and anti-osteogenic TF enrichment, the incipient fetal anlagen (Hypertrophic phenotype) forms at the edges away from the joint, defined by *RUNX2* enrichment (Fig. 2h).

Emergence of fibroblast lineages in synovial joints

Fibroblast lineage cell states have previously been described in the developing mouse limb to arise from a master *HIC1+* precursor population[29], which contributes minimally to osteochondral components. In addition, a "universal" *PI16+* population is thought to persist in postnatal stages and govern adult fibroblast formation across tissues in mice and humans [29,30]. Here, we sought to uncover the taxonomy of the fibroblast lineage in first trimester human joints. We first identified fibroblast progenitors (FibroPro) and *HIC1+* mesenchyme (*HIC1+* Mes) enriched in the appendicular joints during the embryonic period (<8 PCW), surrounding the nascent joint and with diffuse distribution in the limbs, respectively (Fig. 2a, Extended Data Fig.3a-c,f). *TWIST1*, a known activator of postnatal fibrosis and *TWIST2* a regulator of postnatal dermal fibroblast proliferation [31,32], were predicted to show high TF activity in DermFibro (Extended Data Fig. 3e), which was present across sites and time, and localised to the skin around the developing embryonic limbs (Extended Data Fig.3 f-g). Interestingly, *EN1,* a TF required for the fibrotic response and associated scarring during postnatal wound healing[33], was highly expressed in DermFibro (Extended Data Fig. 3e) but showed low target gene expression, suggesting target repression, consistent with observations of scarless wound healing *in utero*. In the cranium, it formed the majority of early-stage enriched fibro-lineage progenitors (Fig. 2a).

To uncover developmental dynamics in the appendicular joints, we reconstructed pseudotime trajectory across fibroblasts and predicted *HIC1+* Mes as a progenitor to FibroPro during the embryonic phase at <8 PCW (Extended Data Fig. 3a). *HIC1+ Mes* enriched for activity in numerous proliferation associated TFs including *WT1, SOX5,* and *FOXC1,* which is associated with an invasive and activated synovial fibroblast phenotype (Extended Data Fig. 3e) [34–37]. At ~8 PCW, FibroPro forms a "fibroblast hub" and expresses *PI16* and *DPT*, markers of pan-tissue adventitia-associated fibroblasts in postnatal health (Extended Data Fig. 3d). Here, it gives rise to tenocytes, synovial, dermal and myo- fibroblasts (Fig. 2a,b). *BNC2,* a myofibroblast-associated TF [38], had high activity in FibroPro, consistent with its postnatal pan-tissue presence in the vascular adventitia [30]. Additionally, *YBX1*, a TF shown to drive proliferation of mouse embryonic fibroblasts was also enriched. *HIC1+* Mes additionally differentiated to the tenogenic lineage during the embryonic, and synovial fibroblasts in the fetal phase, highlighting potential parallel routes to synovial fibroblast formation in the prenatal skeleton (Extended Data Fig. 3a-d).

Formation of the cranium and suture joints

In the cranium, incipient suture mesenchyme matures beyond 7 PCW at meeting points of bone fronts emanating from primary ossification centres in the flat bones, forming suture joints[39]. We conducted differential abundance testing of the mesenchyme compartment and revealed the embryonic cranium (<7 PCW) was dominated by dermal and myo- fibroblasts, and two broad *PAX9+RUNX2+* and *ALX1+RUNX2+* mesenchymal clusters that had not been described previously in human craniogenesis (Fig. 2a). The latter expressed markers such as *TWIST1, CTSK, ZIC1,* consistent with mouse cranial progenitors[40], and *RUNX2*, suggesting they form part of the osteogenic lineage (Extended Data Fig. 4a). *ALX1* is required for cranium formation in the mouse[41], and neural crest cell (NCC) migration and differentiation in human derived stem cells, with variants of the gene associated with frontonasal dysplasia [28,42]. Single-cell studies of the human brain have also identified *ALX1* expression within mesenchyme-like progenitors[43]. We therefore used metadata obtained during tissue sampling to examine differences in anterior-posterior skull gene expression and found numerous neural crest cell (NCC) TFs and targets linked in our GRN including *ALX1, PAX3, BMP5, COL4A2, COL4A1* and *TSHZ2* which are differentially enriched in the anterior portion of the cranium (Extended Data Fig. 4b). This suggests that *ALX1+* Mes may have a neural crest origin. Due to the transient embryonic nature of NCCs[44] we did not capture early embryonic *bona fide SOX10+* NCCs. *RUNX2*-expressing *PAX9+* Mes was present in the cranium (skull base) and appendicular skeleton (Fig. 2a), and therefore likely represents a lateral plate and axial mesoderm-derived progenitor for osteoblasts in endochondral ossification.

Next, we sought to delineate the contribution of the broad ALX1+ and PAX9+ Mes clusters to the bone-forming suture joint. We first sub clustered these populations in combination with osteoblasts (*SP7, ALP*), osteocytes (*SOST, DMP1)* and perichondrium (*RUNX2, THBS2, POSTN)* to form an osteo-lineage subcompartment. This revealed anatomical and age - segregated clusters (Fig. 2a, Extended Data Fig. 4c-d). Among these were three early progenitors, including facial (FacialMes) and pharyngeal mesenchyme (PArchMes) which were located in the skull base and expressed markers of axial mesoderm (*PAX3 and LHX8*), and a cranial mesenchyme (CranialMes) population which was abundant in the calvarium (Fig. 2a, Extended Data Fig. 4d-e).

Cranial sutures form in fetal stages following expansion of the primary ossification centres at the end of the embryonic period[1]. Consistent with this, we define two human suture mesenchyme clusters (SutureMes1/2) enriched in the cranium from 8 PCW (Fig. 2a). Classical markers of fetal cranial sutures (*TWIST1, ZIC1, ZIC4)*, previously reported in NCC-derived mesenchyme in mice, were enriched in both populations in addition to high expression levels

of *THBS2,* akin to the endochondral perichondrium cluster (Extended Data Fig. 4d)[45]. Notably, the SutureMes populations expressed *CTSK*, a marker of periosteal mesenchymal stem cells in the postnatal mouse cranial osteogenic niche shown to mediate intramembranous ossification[46]. Using Cell2location, we revealed the spatial distribution of SutureMes1 and SutureMes2 within the developing coronal suture joint, spanning outward toward the periosteum (Fig. 2i). Osteoprogenitors (*HHIP*+PreOB) emerged at the opposing frontal and parietal bone boundaries of the suture populations (Fig. 2i). Additionally, the sutures were flanked by peri-suture (*CTSK*-) in contrast to the *CTSK*+ postnatal mouse periosteum previously reported[46] (Extended Data Fig. 4g).

Analogous to osteogenic repressor genes enriched in spatially-defined clusters of the articular IZ, SutureMes populations also showed predicted TF and target accessibility enrichment of negative regulators of osteogenesis in the mouse cranial sutures (*TWIST1, LMX1B*, *NFATC2)*[47] (Fig. 2i). Numerous TFs associated with osteogenesis (*EGR1, EGR2, SP7)*[48,49] and osteoblast formation (*FOXO1, ETS1, ETS2*)[50] were predicted suggesting a transcriptional network primed for impending osteogenesis. Akin to the molecular gradients of the joint-interface in the embryonic knee (Fig. 2h), we observed *LMX1B* and *TWIST1* expression within the suture region, dissipating toward the flanking bone edges alongside a rise in regulators of osteogenesis such as *RUNX2* (Fig. 2i), suggesting comparable mechanisms in sustaining the non-osteogenic interface in both the limb and skull bones.

To dissect this process and uncover the enhancer-driven gene regulatory network (eGRN) of the loci surrounding key osteogenic transcription factors *RUNX2* and calvaria-enriched *HHIP,* we visualised coverage of the network predicted by SCENIC+. Both *RUNX2* and *HHIP* were predicted to be inhibited by a shared set of anti-osteogenic TFs including *LMX1B*, *TWIST1* and *ALX4* via several intermediate repressors targeting various enhancers around their loci (Fig. 2k), illustrating the relationships maintaining the balance of osteogenesis initiation. *HHIP* was most accessible in *HHIP*+PreOB and was indirectly repressed by *LMX1B* via *TWIST1* which positively regulated its immediate repressors *MAF* and *HMGA2* via two differentially accessible enhancers of *HHIP* enriched in SutureMes1 and *HHIP*+PreOB. *RUNX2* was most accessible in *HHIP*+PreOB (IM) as well as Perichondrium (EC) and was predicted to be indirectly targeted by the same repressors via *TCF12* and *PRRX2*. Overall, the network illustrates the coherent regulation of bone-adjacent non-ossifying niches by key osteogenic regulators via multiple redundant paths.

Distinct trajectories of osteogenesis across the skeleton

Osteoblastogenesis, signifying the onset of ossification, commences from ~7-8 PCW onward (Fig. 2a). To reconstruct this process, we utilised force-directed embedding on non-cycling cells and uncovered two major trajectories from distinct progenitors converging toward osteoblast and osteocyte formation (Fig. 3a). Endochondral ossification (EC) consisted of two routes, stemming from limb mesenchyme (LimbMes), FacialMes and PArchMes. LimbMes, a cluster which encompasses upper and lower limb joints, demonstrated transcriptional similarity (*ISL1, TBX5, WT1)* to lateral plate mesoderm previously reported in the fetal human limb bud **(**Fig. 3a, Extended Data Fig. 4a-e**)** [11,30]. CranialMes and FacialMes were differentially abundant in the anterior portion of the calvarium and skull base (Fig. 2a), respectively. FacialMes expressed NCC-derived mesenchyme regulators *PAX3* and *PAX7 (*Extended Data Fig. 4e)[51]. Likewise, *ALX1* was enriched in CranialMes, leading us to hypothesise that the clusters represent, to our knowledge, previously undescribed human NCC-derived osteogenic populations. *LEPR*+Mes was enriched in the calvarium, *LEPR* has been reported as a marker of second trimester and post-natal osteogenic stromal population [12,52]. PArchMes emerged between 6-8 PCW and expressed marker genes of the first pharyngeal arch (*PAX9, LHX, DLK1) *[53], and is likely to comprise mixed origins from axial mesoderm and NCC. We discovered a two-stage process in the intramembranous pathway (IM), defined by the transition between embryonic and fetal stages. Suturogenesis occurs in stage 1, as CranialMes differentiates into SutureMes1/2 (Fig. 3a). In stage 2, an osteogenic trajectory emerges from SutureMes1 toward SutureMes2, followed by *HHIP*+PreOB. *HHIP* was previously reported in the mouse as a marker of osteogenic coronal suture mesenchyme [54]. We demonstrate here that they are a distinct population progeny to *TWIST1-enriched* SutureMes and are distributed at the boundaries of bone formation in human fetal development (Fig. 2i).

Within the two newly defined trajectories for EC and IM osteogenesis, we determined dynamic expression and modules associated with pseudotime as they converge (Fig. 3a-b). Common TFs (*RUNX2, DLX5, SP7, SATB2)* mediating osteogenesis were upregulated toward terminal states across both pathways. The newly described SutureMes1/2 was enriched for osteogenic genes (*SPARC, COL1A1, GAS1)* early in the trajectory and signatures associated with suture formation and regulation in the fetal and postnatal mouse suture (*PRDM6, MSX1)* were most regulated upon transition to SutureMes2 [55]. Inhibitory control of osteogenesis differed across the pathways, *LMX1B* a common repressor, was enriched late in progenitors across both pathways whereas *TWIST1* was only enriched late in IM progenitors, signifying additional inhibitory regulation in the IM pathway. *NFATC2*, a chondro- and osteo- genesis repressor

[56,57], was observed early in the IM pathway, but upregulated late in the EC pathway, potentially reflecting context and timing dependent roles in maintenance of the suture niche (IM), and cartilage primordia replacement (EC), respectively. In IM progenitors, accessibility across activators (*RUNX2*) and repressors (*TWIST1, LMX1B*) was simultaneously high at early parts of the trajectory, suggesting they are poised for osteogenesis but remain repressed in the early stages (Fig. 3c). In contrast, at the transcriptome level, there was a reciprocal relationship between repressor and *RUNX2* target expression across pseudotime, suggesting additional, redundant layers of regulation limiting RUNX2 target expression. In the EC pseudo-trajectory, repressor (*TWIST1, NFATC2*) accessibility peaked toward Perichondrium and then decreased, whereas *RUNX2* target accessibility increased as Perichondrium formed, suggesting the latter was critical in driving progression. This contrasts with the gradual reduction in IM repressor accessibility marking the transition into *HHIP*+PreOB alongside a persistently accessible activator (*RUNX2)* (Fig. 3c). In concert, the multi-omic dynamic changes in the IM pathway, in contrast to EC osteogenesis, suggest additional layers of non-transcriptional regulation.

Following suture formation, progressive waves of oriented differentiation emanate from the cranial sutures toward the developing bone front [45]. In order to identify the newly defined cell states in this progression (stage 2), we utilise OrganAxis (see Methods) to define a maturation axis spanning the coronal suture into regions of the maturing frontal bone. Subsequently, we create zonal bins within this continuous axis based on histological features, and evaluate cell-state mapping along the anterior-posterior (AP) axis (Fig. 3d). Enrichment of *TWIST1*+ SutureMes1/2 were substantial in the suture zones (1-3) which constitute the centre of the coronal suture (Fig. 3e). Their enrichment peaked within the osteogenic front where histological features of osteoprogenitors began to emerge along with *HHIP*+ PreOB. Establishment of the osteogenic zones coincided with downregulation of anti-osteogenic (*LMX1B, TWIST1)*, and upregulation of pro-osteogenic (*RUNX2, DLX5, SP7)* TFs (Fig. 2i-j, 3d-e), signifying a spatially restricted molecular switch that defines territories of the suture. Osteoblast-specific genes including *IRX5, SOST, SPP1, MMP9 and DMP1* gradually peaked toward the most distant osteogenic zone[58]. Clustering of the spatial voxels aligned osteogenic TFs with axis values away from the coronal suture (Extended Data Fig. 5c).

In EC ossification, bone formation initiates in the interfaces of the synovial joint, with perichondrium formation surrounding the hypertrophic chondrocyte template, giving rise to osteoblast in fetal stages and in postnatal injury repair [11,52,59]. At the border of the sphenoid in the anterior skull base, we identified the anterior boundary between EC and IM osteogenesis

at the frontal-sphenoid interface, with co-localisation of SutureMes1 and SutureMes2, which

precede *HHIP*+PreOB and Osteoblast (Fig. 3e), forming a previously undescribed boundary-

defining niche. In the limb, we applied ISS-patcher to the ISS data, and inferred spatial

trajectories for differentiation of the LimbMes towards perichondrium and osteoblasts in a

centripetal pattern in the incipient bone at ~7 PCW (Fig. 3f). In the older 9 PCW skull base, a

comparable pattern was detected in the sphenoid where hypertrophic chondrocytes were

surrounded by perichondrium (Fig. 3g).

To further resolve the endothelial-bone relationship within the IM niche, which remains poorly

understood compared to the canonical vascularisation models described in the EC niches[60],

we leveraged both the spatial locations and inferred osteo-endothelial interactions to elucidate

a plausible cellular basis for endothelial sprouting. As expected within the IM niche, $MSR1^+$

Tip cells, Mural cells and Capillary Endothelial cells (ECs) progressively co-enriched along the

osteogenic zones with Osteoblast and Osteocytes (Fig. 3e, Extended Data Fig. 5a-b),

suggesting a spatial-temporal association between vascularisation and osteogenesis.

Concurrent to this trend, *ATF4*, a regulatory gene that promotes bone angiogenesis in

development [61], was also enriched along the axis.

Pro-angiogenic factors *VEGFA* and *VEGFB,* the concentration of which modulates vessel

sprouting, were highly enriched at the osteogenic front, which co-localised with SutureMes1/2

and $HHIP^+$PreOB, which highly expressed *VEGFA* and *VEGFB*, suggesting they are the likely

source. Endothelial populations highly express the *VEGF* receptor genes (*FLT1, KDR, NPR1*)

(Fig. 3h), which therefore suggests the former plays a pivotal role in promoting vascular

invasion into the bone in IM akin to the well-described function of hypertrophic chondrocytes

in EC niches[62,63].

To determine distinct cell-cell interactions across the EC and IM niches, we utilised NicheNet

to compare differences in signalling during sprouting angiogenesis, focusing on recipient tip

cells that lead the vascular sprouts. SutureMes1/2 distinctly enriched for *EPHB2* and *EPHB3*,

which promotes sprouting behaviour, motility, and vessel formation (Fig. 3i) [64]. *RACK1*, an

intracellular scaffold protein that promotes *VEGF-FLT1*-dependent cell migration[65], was

exclusively expressed in tip cells in the calvaria, signifying a highly motile state. *CXCR4*, which

promotes sprout anastomosis, was upregulated in tip cells[66]. In EC ossification, tip cells

distinctly express *CDON*, the receptor for *IHH* which promotes endothelial proliferation,

migration, and angiogenesis. Together, these data support the essential role of SutureMes1/2

and $HHIP^+$PreOB in coordinating blood vessel invasion and migration in the IM osteogenic

niche.

Next, we sought to unravel inferred signalling interactions from the invading endothelial cells
toward osteogenic populations. We first defined a spatial gradient of angiogenesis along the
defined bone-maturation axis by scoring for enrichment in sprouting angiogenesis pathways
(Fig. 3e), which corroborated the coupling of angiogenesis and osteogenesis, orchestrated in
the direction of bone growth. We then used CellphoneDB to predict significant signalling
interactions from ligands in the endothelial populations toward the recipient osteogenic cell
states (Fig. 3j)[67]. Endothelial cells, including capillary ECs, mural cells and tip cells were
predicted to signal to osteoblastic cells via *DLL1/DLL4–NOTCH2* and *JAG1/JAG2-NOTCH2*
interacting pairs, which have been reported to promote the differentiation of postnatal
perivascular osteoprogenitors[68]. Of note, no *NOG* expression was observed in endothelial
cells, unlike the previously described postnatal bone-associated type H vessels[68], which
suggests a novel and unique pattern of crosstalk between endothelial and osteoblastic cells
via *NOTCH* signalling in the embryo. Additionally, mural and capillary ECs promoted
osteoblast differentiation through previously described interactions including *FGF2-FGFR2*[69]
and *THBS1-CD36*[70]. Mural cells are also likely to be promoting bone mineralization via
activation of canonical *WNT* signalling through the observed RSPO3-LGR5 pairs[71]. In addition
to pro-osteogenic roles, the endothelial cells all support osteoclast recruitment and
differentiation via *CCL14-CCR1*[72] and *CXCL12-DDP4* interactions[73]. Through these spatially
defined interactions, we discern a model of the interplay between angiogenesis and
osteogenesis in the intramembranous niche of the frontal bone whereby tip cells are first
recruited by *VEGF* and *EPHB2* from suture mesenchyme and pre-osteoblastic cells, toward
the pre-osteogenic condensation that form the bone-front. Vascularising endothelial cells then
promote osteoblastic differentiation, osteocyte mineralization and osteoclast recruitment in the
maturing bone (Extended Data Fig. 5e).

Other lineages, including neurons have also been reported to play a role in modulating
osteogenesis[74,75]. Neuronal development and axon guidance modules were enriched in early
stages of both osteogenic pathways (Supplementary Table 4), suggesting involvement of
chemical cues from axons in niche formation. Specific IM modules were associated with
epithelial mesenchymal transition (EMT), hypoxia response, and TGF signalling, whereas EC-
enriched modules were related to neuroactive ligand-receptor interactions, axon guidance and
development (GSEA, MSigDB Hallmark 2020 gene sets). To understand the potential role of
axon guidance in skull formation, we compute gene-scores for axon guidance modules along
the spatial trajectories of osteoblastogenesis in the cranium and revealed enrichment in the
suture mesenchyme (Extended Data Fig. 5f), consistent with previous description of
axonogenesis regulating suture formation in the mouse[76]. Future studies capturing

transcriptional profiles of neuronal cell bodies of the innervating axons may shed light on the
cell-cell interactions at play.

Inferring *PAX7*+ chondrocyte origins through somatic mutations

Chondrogenesis is an essential process for the formation of various types of cartilage,
including hyaline, fibrous, and elastic cartilage. To reveal the transcriptional heterogeneity of
chondrocytes, the chondrocyte subcompartment was further sub-clustered across anatomical
regions and timepoints and annotated based on canonical marker genes (Fig. 4a, b, Extended
Data Fig. 6a and Supplementary Table 2). To dissect cellular heterogeneity within a spatial
context, we applied ISS-Patcher and Cell2location to transfer cell state labels from our single-
cell atlas to ISS and 10x Visium slides from the shoulder at 7.3 PCW and calvaria at 9 PCW
(Extended Data Figs. 6b, c).

Our chondrocyte clusters exhibited strong regional-specificity due to divergent development
origins (Fig. 4c)[77]. In terms of cell clusters, we identified known populations, such as
hypertrophic chondrocyte (HyperChon, *COL10A1*, *IHH*), cycling chondrocyte (CyclingChon,
MKI67, *TOP2A*), resting chondrocyte (RestingChon, *UCMA*), interzone chondrocyte
(InterzoneChon, *GDF5, PITX1*), as well as new populations, including two chondrocyte
progenitors (ChondroPro1, ChondroPro2), anatomical region-specific chondrocytes, *DLK1*+
chondrocytes (*DLK1*+ Chon, *DLK1*, *CD63*), and *PAX7*+ chondrocytes (*PAX7*+ Chon, *PAX7*,
RBFOX1). Two region-specific ChondroPros, which were enriched in appendicular joints and
skull respectively, transitionally expressed fibroblast differentiation markers (*POSTN*,
COL1A1, *PRRX1*, and *TWIST1*), which was consistent with findings in early chondrocyte
progenitors in mice[19]. In the skull base, facial chondrocytes highly express *PAX3*, indicating
an origin from neural crest[78]. Mandibular chondrocytes highly expressed *SEMA3D* in posterior
regions[79,80]. In appendicular regions, we observed two subtypes of articular chondrocytes, with
one population being enriched for *TRPV4* and *VEGFA*, while the other one was more mature,
with relatively low *SOX9* expression and high *EPYC* expression. In particular, *DLK1*+ Chon
was highly enriched in ribosomal genes and distinctly expressed *CD63* and *DLK1*. *CD63* has
been identified in the pre-hypertrophic layer in the limb, while *DLK1* is a novel marker for
embryonic chondroprogenitor cells undergoing lineage progression from proliferation to pre-
hypertrophic stages. Spatially, CyclingChon, *DLK1*+ Chon and HyperChon were also arranged
sequentially within the bone, spanning from the epiphysis toward to diaphysis, the primary
ossification centre (Extended Data Fig. 6b), which corroborates the transitional role of *DLK1*+
Chon, from proliferative to pre-hypertrophic phenotypes during chondrogenesis.

With regard to gene regulatory networks, we identified common and region-specific co-
expressed gene modules using Hotspot[81] and performed gene ontology enrichment on each
module (Extended Data Figs. 6 d,e). Shared gene modules (M3, M5, etc) across regions
mainly related to chondrocyte cellular structure and functional development, including
biomineralization, collagen-containing extracellular matrix, and cartilage development.
Interestingly, synapse related pathways were enriched both in appendicular and cranial
samples, suggesting axons may contribute to development of cartilage[82] which in pathology
may be responsible for joint pain. In addition, chondrocytes from the skull base enriched for
axon guidance gene modules, while chondrocytes from appendicular joints had specific gene
modules, including glucose metabolism and cellular response to hypoxia (Extended Data Figs.
6f, g). Taken together, our data provides a comprehensive catalogue of chondrocyte molecular
profiles in multiple developing human joints.

Moreover, our work uncovered a previously undescribed developmental chondrocyte cluster
(*PAX7*+ Chon), which first appeared at 7 PCW and peaked between 9-10 PCW across multiple
donors (Fig. 4c). *PAX7*+ Chon co-expressed markers and gene modules of chondrocytes and
muscle cells on the transcriptomic and epigenetic level (Fig. 4d), and showed overlap of genes
that enrich in cartilage and muscle development pathways (Extended Data Fig. 6h, i). To
interrogate the possibility of doublet-capture artefact, we first applied stringent doublet
processing by performing extra statistical tests over the doublet score (see Methods, Extended
Data Fig. 7a, b). Then we verified the presence of *PAX7*+ Chon through *in situ* hybridization
experiments (Fig. 4e), which was consistent with the multiplexed ISS data (Extended Data
Fig. 6b). Additionally, *PAX7*+ Chon displayed transcriptional characteristics not explained by
doublet formation (Extended Data Fig. 7b,c). To reveal GRNs that govern the transcriptional
identity of *PAX7*+ Chon, we applied SCENIC+ and constructed a GRN based on differentially
expressed transcription factors (TFs) with their predicted target genes. Aside from identifying
key myogenic and chondrogenic key regulators (*PAX7*, *MYF5*, *SOX5*, *SOX9*), we also
discovered core posterior axis HOX gene modules, suggesting that patterning of major
synovial joint chondrocytes may be conferred through their nested HOX expression [83].

Combined profiling of RNA and ATAC offers deep coverage for mutation detection at the single
cell level. Under the assumption that shared somatic mutations should be observed in cells
derived from common progenitors, we used SComatic[84] (see Methods) to detect somatic
mutations in the form of single-nucleotide variants (SNVs) at the single cell resolution to
ascertain whether chondrocytes may share origins with the muscle lineage, as suggested by
previous mouse studies[85,86,87]. We identified shared mutations between *PAX7*+ Chon and

myogenic cell states across multiple donors (Fig. 4g, h). The latter included three progenitor

cell states. Skeletal muscle mesenchymal cells (SkM.Mesen)[15] not only express classical

muscle markers at low levels, but also exhibit gene expression related to mesenchyme. Myo-

skeletal stem-like cells (MSSCs) expressed *CADM1*, *MECOM*, and *FOXP1*, previously

reported as makers of mouse skeletal/stem progenitors[88], along with genes related to muscle,

cartilage, and mesenchyme. In addition, MSSCs expressed low levels of lateral plate

mesoderm markers. Conversely, *PAX7*+ MyoProgenitor cells highly expressed muscle-related

genes, suggesting they are a distinct branch separate from *PAX7*+ Chon (Extended Data Fig.

7c). We further interrogated the relationship between these progenitors and *PAX7*+ Chon,

applying scFate, to reveal a trajectory from MSSCs to *PAX7*+ Chon, with upregulation of

chondrocyte genes and downregulation of stem cell genes, such as *SOX4*, *CDH2* and *CDH11*

on both the RNA and ATAC level (Extended Data Figs. 7d-f). Comparison of mouse and

human cell clusters revealed similarity between *PAX7*+ Chon and a *MYF5*+ progenitor-derived

mouse cluster from the extraocular muscle (Extended Data Fig. 7g), which was reported to

derive from progenitors that give rise to both myogenic and connective tissues in mouse[15,87].

Taken together these results indicate the possibility of a new source of chondrocytes from

muscle progenitors, which suggested the potential complexity of embryonic cell types prior to

reaching terminal differentiation.

Lineage tracing experiments in zebrafish and mice[89] have shown that Schwann cells possess

the evolutionarily conserved potential to differentiate into chondrocytes during embryonic

development in both species. Here, we identified SOX9+ endoneurial Schwann cells (SOX9+

enSC) within the Schwann cell compartment, characterised by the expression of typical

chondrocyte (*SOX9*, *COL9A1*, *ACAN*, *COL2A1*), and classical Schwann cell markers (*MPZ*,

SOX10) (Extended Data Figs. 8a-c). Velocity analysis suggested that multipotent hub

Schwann cells (Hub SCP) serve as the foundational cells in the differentiation hierarchy.

SOX9+ enSC represented one of the endpoints on the mesenchymal trajectory, expressed a

mesenchymal gene signature (*PRRX1*, *PRRX2*, *PDGFRA*, *TWIST2*) and upregulated HOX

TFs (*HOXA9*, *HOXA10*, *HOXA11*, *HOXD10*) (Extended Data Figs. 8d, e). We applied

RNAscope (*SOX9*, *SOX10*, *MPZ*) to localise SOX9+ EnSC in the acetabulum at 7.3PCW

(Extended Data Figs. 8g, h). Shared mutations between chondro-lineages and Schwann

lineages were enriched in SOX9+ enSC, indicating that SOX9+ enSC exhibit a close

differentiation relationship with chondrocytes (Extended Data Figs. 8i). Most of these

mutations were concentrated in the skull base, consistent with the process described in the

craniofacial and trunk skeleton of mice[89]. Notably, a limited number of mutations were also

detected in the hip, consistent with the findings from the RNAscope experiments. In summary,

the analyses suggest that Schwann lineage cells may carry the potential to serve as an

alternative source of chondrocytes in humans. However, further lineage tracing experiments
are essential to validate this. We therefore hypothesise that chondrocytes derive from multiple
potential origins, such as muscle and Schwann lineages during embryonic development (Fig.
4i).

Developmental links to complex traits of the adult skeleton

Numerous conditions of the postnatal and ageing skeleton have been linked to disrupted joint
and bone changes during the embryonic stages of life. Of note, enhancer-associated variants
linked to adult osteoarthritis (OA) appear to act on anatomical-region-specific regulatory
networks to influence synovial joint morphology during development[90]. However, the cell
states that mediate this process remain evasive.

To explore potential causal links in foetal skeletal cell states for adult complex traits, we first
conducted fGWAS (Fig. 1c, see Methods), which integrates full GWAS summary statistics with
cell-state specific epigenetic and transcriptomic signatures across clusters, computing cell-
cluster enrichments for phenotypes. Interestingly, we found distinct lineage enrichments for
knee and hip OA in chondrogenic and osteogenic cell populations, respectively (Fig. 5a). Knee
OA and its surrogate phenotype, total knee replacement (TKR) enriched in most chondrogenic
states, except for $PAX7^+$ chondrocytes and InterzoneChon ($GDF5^+$). In contrast, hip OA
enrichment was only significantly observed in two chondrocyte populations (ChondroPro and
Hypertrophic Chondro), but was additionally enriched in Perichondrium, Osteoblast as well as
Tenocyte. We further performed fGWAS on our fine-grained annotations (Fig. 5b) to explore
subcluster enrichment. Consistent with the broad annotations, various types of chondrocytes
show significant enrichments for knee OA (Fig. 5b) whereas the signal for InterzoneChon
implicated *Early* IZ in hip OA. These findings implicate a greater impact for bone development
affecting hip OA risk, potentially through modulating hip shape and a thereby altered risk of
mechanical stress at the hip joint over time[91]. In contrast, the knee-specific findings point
toward the central role of chondrogenesis in knee pathology, potentially implicating postnatal
regenerative mechanisms.

Complex diseases are often associated with many mutations across non-coding regions. We
therefore sought after a way to integrate the GWAS signals of hip and knee OA described
above across enhancers of our gene regulatory network to identify enriched TFs and
pathways, while accounting for cell cluster specific effects. To this end we developed
SNP2CELL, a tool that uses gene regulatory networks as a basis to aggregate scores of
individual SNPs and cluster marker genes across pathways of connected TFs (Fig. 5c).

Aggregation of scores was achieved through a network propagation approach[92]. In contrast to

simply linking SNPs with their closest genes, this also allows for the identification of upstream

TFs which might not be mutated, but whose effect on multiple binding sites and target genes

is often disrupted, by integrating a large number of disease associations. By using eGRNs

previously computed with SCENIC+, we can utilise SNP signals mapped to both gene regions

and distal enhancers in ATAC peaks. By overlapping integrated SNP scores with integrated

marker gene scores per cell cluster, and comparing the effects to random permutations, we

obtain cell cluster specific enrichments of sub-networks of our initial eGRN.

Deriving insights from cell states implicated in fGWAS, we next clustered enrichment scores

for knee and hip OA, focusing on osteogenesis for the latter. In knee OA, various chondrocyte

subtypes clustered together by high enrichment score and average gene enrichments (Fig.

5d), mirroring its fGWAS enrichment for chondrocytes. Likewise for hip OA, Perichondrium

expectedly had the largest average enrichment across the osteogenic trajectory (Fig. 5e)

consistent with fGWAS results. Utilising SNP2CELL, we identified sub-networks for knee OA

and hip OA, prioritising articular chondrocytes (Fig. 5f) and perichondrium (Fig. 5g),

respectively, and revealing similar regulatory pathways that balance chondro and osteogenic

functions.

In the knee articular chondrocyte knee OA network (Fig. 5f), a number of non-TF genes with

roles in cartilage makeup and chondrocyte differentiation (*COL27A1, PRKCA, SNORC,*

CRISPLD2) were identified and predicted to be regulated by *NFATC1* and *FOXA3*. *COL27A1*,

and the proteoglycan *SNORC*, are involved in chondrocyte ECM makeup and maturation[93].

However, *PRKCA* is a kinase that has been linked to mechano-sensing in articular

chondrocytes[94]. Notably, *NFATC1* itself has been described as a marker of articular cartilage

progenitors affecting differentiation[95] and also protects against OA[96]. For hip perichondrium in

hip OA (Fig. 5g), the osteogenic key regulator *RUNX2* showed significant enrichment, along

with multiple *NFAT* genes (*NFATC1, NFATC2, NFATC4, NFAT5*), which in conjunction with

additional TFs (*ZEB1, MAF, TEAD1*) implicated calcineurin and *WNT* signalling pathways.

Both canonical and non-canonical *WNT* signalling via Ca^{2+}/Calcineurin/*NFAT* have been

linked to bone formation and remodelling, affecting the balance and differentiation of both

osteoblasts and osteoclasts [97]. Various *WNT* signalling inhibitors including *DKK1* and *FRZB*

have also been linked to the shaping process of the hip and osteoarthritis, which may be

particularly important during foetal development[98]. To provide a summary of the potential

functional role of the enriched disease-associated cell-cluster specific genes, we conducted

gene set enrichment analysis (GSEA) on Gene Ontology (GO) terms. This expectedly featured

terms related to extracellular matrix organisation, cartilage development and chondrocyte

differentiation in articular chondrocytes in knee OA specifically, whereas hip perichondrium

scores showed enrichment for inositol-phosphate, calcineurin and *NFAT* signalling, and

cellular response to lipid (Fig. 5h). The latter points to the potential interplay of lipids with bone

formation and osteoarthritis[99]. Overall, through application of fGWAS and our newly developed

pipeline SNP2Cell, we were able to distil differential involvement of the chondrogenic and

osteogenic lineages respectively in hip and knee OA, and provide a cellular basis for the link

between developmental cell states and regionally distinct adult disease.

Deciphering monogenic conditions of the embryonic skeleton

Both syndromic and non-syndromic forms of craniosynostosis are congenital conditions that

involve disturbances in cranial ossification and suture formation during fetal and postnatal

development. Disease-associated clinical features include premature cranial suture fusion,

and depletion of osteoprogenitor pools that facilitate the postnatal expansion of the cranium,

leading to severe global developmental consequences. Disease mechanisms that underlie

these changes are reportedly linked to missense mutations and haploinsufficiency in genes

that govern persistence of IM osteo-progenitor pools within the suture joints throughout the

cranium [100–102].

To implicate cell states that underlie craniosynostosis disease mechanisms we first cross-

referenced pseudotime-associated DEGs enriched in the IM and EC pathways against a

candidate database of over 2700 genes known to cause congenital conditions in humans[11].

We then filtered for enriched DEGs that were associated with musculoskeletal (MSK)

conditions and created a global view of enriched expression and accessibility across the

intersecting genes (Fig. 6a). Within these DEGs, we then filtered for craniosynostosis

associated genes by combining a subset of the database, and reported genes obtained from

Genomics England Limited rare and common craniosynostosis panels (Supplementary Table

5), of which a high proportion of the top DEGs were associated with synostosis (Fig. 6a). Most

of these associations (n=13/22) were within the progenitor populations of the IM trajectory

(SutureMes1/2 and *HHIP*+PreOB). Broadly, most of the craniosynostosis enriched genes

were highly accessible and expressed in IM progenitors, apart from *FN1 and IHH*, suggesting

the embryonic period is critically affected in craniosynostosis.

To elucidate potential effects of the synostosis-associated TFs on gene changes during

osteogenesis, we applied CellOracle to predict velocity shifts within the newly defined

trajectory and eGRNs using *in silico* perturbation simulation of gene knockouts in the IM

osteogenesis sub-compartment (Fig. 6b-c). Of the top-genes predicted to cause greatest

velocity shifts in the trajectory, *TWIST1, MSX2, LMX1B* were all disease-associated genes

that lead to highest velocity shifts in SutureMes2 when perturbed (Fig. 6c). The former two
were also IM pseudotime-associated DEGs (Fig. 6a), suggesting the velocity shift is highly
likely to reflect the loss-of-function mechanism of disease pathogenesis in these TFs [103,104].
Visualisation of the inferred velocity shifts revealed expected patterns of knockout favouring
lineage progression and bone formation (Fig. 6d), which are likely to penetrate in regions of
the coronal suture that indeed correspond to spatial enrichment of the candidate TFs (Fig. 6e).
Numerous other disease-associated TFs demonstrated maximal enrichment in the Suture
Zones (*IL11RA, SIX1*), whereas others preferentially affected more mature parts of the
trajectory (*IHH, ALPL, VLDLR*) (Fig. 6f). Overall, these findings constitute novel explanations
for features of TF-mediated craniosynostosis including premature suture fusion and
ossification.

To explore potential cell-extrinsic influences on fetal bone development, we curated a list of
65 clinically approved drugs from the chEMBL database which carried warnings of
teratogenicity (Supplementary Table. 6). We then utilised to score target gene-group
expression per cell state within our osteogenic sub compartment. Interestingly, this revealed
overall greater target enrichment of teratogenic drugs within IM progenitors and downstream
osteoblast/osteocyte cell states compared to EC progenitors (Extended Data Fig. 9a). Retinoid
drugs expectedly enriched for targets in Osteocytes, consistent with observations of retinoic
acid enhancing *in vitro* osteogenesis in mouse suture cells[105]. Likewise, the well-known
skeletal development teratogen thalidomide enriched for targets in Osteocytes. Aside from
antihypertensives targeting the renin-angiotensin pathway (ACEi, ARB, Renin inhibitor) which
enriched for targets in SutureMes1 and *HHIP*+PreOB, we also identified high enrichment for
endothelin receptor antagonists (ERAs) specifically in SutureMes1. Mice lacking the
endothelin A receptor recapitulate wide-spread cranial neural crest-related defects *in utero*[106].
Informed by teratogenic effects of ERA use in animals[107], they are not routinely prescribed for
pulmonary arterial hypertension (PAH) during pregnancy. However, there remains a lack of
evidence for teratogenicity in humans, and women of child-bearing age continue to utilise them
for treatment of PAH. Our analyses support the notion of potential targets for ERAs within the
human suture cells and lend support to a contraindication for their use during pregnancy.

To further resolve the regulatory relationship between the prioritised TFs (*TWIST1, MSX2,*
LMX1B) within the osteo-lineage, we utilised our paired snRNA-seq and scATAC-seq dataset
to reconstruct interaction networks across the TFs (Fig. 6g). Our predictions revealed a rich
web of inter-regulation of shared coding and non-coding targets across the three candidate
TFs. Of these connected nodes, numerous TFs are associated with loss-of-function mutations
leading to craniosynostosis (*SIX1, TCF12, NFIX, ALX4*), suggesting a tightly regulated

network conferring suture patency, inhibiting osteogenesis in this region. Notably, *TCF12* was reported to govern normal coronal suture development through heterodimer formation with *TWIST1*, and severe phenotypes are observed in mice with doubly deleterious mutations[103]. *TCF12* served as a node connecting *TWIST1* to *RUNX2*, suggesting that disruption of this critical co-regulatory network leads to human disease. While direct transcriptional regulations of *RUNX2* by the three inhibitory TFs were predicted to be weaker than their stronger inter-regulation and not visualised in Fig. 6g, predicted regulations of *RUNX2* (Extended Data Fig. 9b) therefore likely contribute to known complex regulatory mechanisms, including protein-level interactions[108]. In addition, cell-extrinsic signals and control mechanisms are likely to act on the connecting nodal points such as *NFATC2* [97], governing the balance of osteogenesis.

Multi-omic information additionally allows exploration of enhancer-mediated regulation of transcription. This is of particular importance for monogenic bone-forming conditions such as Van Buchem disease (VBD) whereby mutation in the non-coding *ECR5* enhancer of *SOST* leads to sclerosing dysplasia of bone[109]. We therefore constructed a regulatory network centred around *SOST* and identified the *ECR5* enhancer to directly regulate *SOST* (Fig. 6h). Numerous osteogenesis-associated TFs were predicted to regulate the region containing the ECR5 enhancer (*DLX5, KLF2, KLF4, KLF13, MEF2C)*. Notably, *MEF2C* has previously been shown to regulate *ECR5* in mice[110], confirming its direct role in controlling *SOST* transcription. Overall, these predictions form a basis for constructing cellular models of TF and enhancer-driven monogenic conditions of the bone lineage.

Discussion

We present the first multiomic cell atlas to capture the spatially-resolved cellular composition of bone, synovial and suture joint development across the first trimester in humans. First, we dissected the cellular taxonomy of interzone cells, including *GDF5*-expressing progenitors in the synovial joints, and developed ISS-Patcher, a tool to aid inference of IZ subclusters in our high-resolution 155-plex ISS data. We highlighted Articular, Fibro (*GDF5+*) and (*GDF5-*) Hypertrophic IZ subclusters which zonate the articular chondrocyte and cartilage template-forming boundaries of the embryonic synovial joints, respectively.

Within lineages described in mice, *GDF5+* cells are posited to give rise to the numerous cell components within the joint including ligaments and tendons. Here we describe a novel two-stage process whereby the fibroblasts and tenocytes of the synovial joints arise from embryonic *HIC1+* fibroblasts, a population previously reported in the mouse embryonic limb [29], and a fetal (*PI16+*) fibroblast progenitor which is transcriptionally similar to human postnatal

universal fibroblasts [30]. Together, these shed new light on the cellular architecture of the developing human embryonic synovial joint in the first trimester.

We have identified numerous previously undescribed human cranial embryonic osteo-lineage cell states including CranialMes (*HAND2*)[111,112], FacialMes (*PAX3*)[51] and PArchMes (*LHX8*)[113], by their expression of comparable marker genes in mice counterparts. We leveraged our spatial analyses of the fetal cranium to define two novel fetal *TWIST1*[+] SutureMes populations, and *HHIP*-expressing pre-osteoblast which contributes to osteogenesis, and mirrors populations of the fetal mouse suture [30,54]. We also discovered distinct IM and EC osteogenic cellular trajectories in their respective regional niches. Using *OrganAxis*, we illustrate the spatial component of these trajectories as they develop and show a regulatory shift in the poised osteogenic suture population as it progresses in space toward maturing pre-osteoblastic cell states where we uncover a region that critically recruits endothelial cell states to drive ossification. We show that the EC niche in the appendicular anlagen derives from an undifferentiated progenitor expressing *ISL1* and *TBX5*, which mark an equivalent population in the human limb [11], and proceeds to form perichondrium and osteoblasts through developmental time and space in the first trimester. These novel cellular trajectories form a valuable reference for future studies of bone development. Further studies of the second trimester, when adipose cell states and mature ligamentous tissues develop, will facilitate a complete census of the joint.

Through our application of a modified workflow to conduct somatic mutation tracing within our diverse chondrocyte subcompartment, we lend support to a potential myoprogenitor-related origin for the newly reported *PAX7*[+] chondrocyte, and likewise suggest that a novel *SOX9*[+] Schwann cell state may confer chondrogenic potential. These proposed cross-lineage cell fates uncover potential similarities to mouse development described above, and crucially, the value of computational tool development in unravelling comparative human biology. In the future, more focused investigations that involve isolation of these cell lineages will further inform their biology.

Leveraging our dataset, we were able to identify developmental links to both monogenic and complex musculoskeletal diseases. We observed marked differences in the enrichment of hip and knee OA GWAS signals across cell clusters, implicating bone formation in the former, and chondrogenesis in the latter, which we traced back to the osteogenic perichondrium, and numerous chondrogenic cell states, respectively. In future, this framework may be applied to other conditions to gain insights into cell-state specific pathogenesis. Lastly, by systematically simulating *in silico* KO of TFs known to be associated with craniosynostosis, we identified a

network of regulators inhibiting osteogenesis, which provides mechanistic detail of how previously reported monogenic loss-of-function mutations in, for example, *TWIST1, MSX2 and LMX1B,* may lead to early suture fusion. This approach, applied to the newly mapped cross-region osteogenic trajectory, holds immense value in predicting potential effects of gene changes across other diseases involving the cellular lineage.

Our comprehensive multimodal human fetal skeletal atlas is a unique fundamental resource for the understanding of human cartilage and bone development in the first trimester. It also has the potential to serve as a cellular roadmap for *in vitro* endeavours to differentiate osteoblast and chondrocyte and mesenchymal cell states in the developing musculoskeletal system.

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Author contributions

K.T., L.F. AND J.P.P. conceived the study, performed data generation and experiments, conceived methodology, performed analysis, collected biological samples, and wrote the paper. K.R. performed data generation, conceived methodology, collected biological samples, performed analysis, conducted imaging, and wrote the manuscript. T.L., N.Y. performed computational analysis and conceived methodology. P.H. performed computational analysis and provided key biological interpretation. C.X. performed computational analysis and provided key biological interpretation. J.C. performed experiments, conceived methodology, and collected samples. R.L.. performed computational analysis and provided key biological interpretation. K.K. performed computational analysis and provided key biological interpretation. N.H. performed analysis and provided key biological interpretation. S.M. performed analysis and provided key biological interpretation. L.R., R.K., S.P., A.W.C. performed experiments and sample processing and collection, generating single-cell data. E.T, I.M. and F.M. performed imaging experiments, provided key biological insights, and collected tissue samples, and generated spatial data. B.C., A.V.P., D.H., S.M., M.P. and P.M. performed computational processing of data and performed data analysis. X.H. conducted tissue processing and sample collection. K.B.M. conceived computational methodology and provided key biological interpretations. M.H. conceived the study and provided key biological interpretations. R.A.B. conceived the study and provided key biological interpretations. O.B.

conceived the imaging experiments, provided key biological interpretations and analysed the data. C.D.B. conceived the study, conceived methodology, provided key biological interpretation, and supervised the project. S.A.T. conceived the study, conceived methodology, provided key biological interpretation, and supervised the project.

Conflict of interest

C.D.B is a founder of Mestag Therapeutics. S.A.T. has consulted for or been a member of scientific advisory boards at Qiagen, Sanofi, GlaxoSmithKline, and ForeSite Labs. She is a consultant and equity holder for TransitionBio and EnsoCell. The remaining authors declare no competing interests.

Data availability

High-throughput raw sequencing data in this study is available from ArrayExpress (www.ebi.ac.uk/arrayexpress) with accession numbers. Processed snRNA-seq/scATAC-seq data, visium and ISS data are available for visualisation and downloading from **https://developmental.cellatlas.io/skeleton-development**

Code availability

ISS-Patcher is available at **https://github.com/Teichlab/iss_patcher .**
SNP2CELL is available at **https://github.com/Teichlab/snp2cell.**

Materials and Methods

Sample acquisition and ethics
Developing human limb and cranium tissue samples were obtained from elective terminations under REC 96/085 (East of England - Cambridge Central Research Ethics Committee). Briefly, samples were kept suspended in PBS and at -4°C on ice during dissection. Shoulder, hip and knee joints were dissected en-bloc from the limbs. For the shoulder joint a proximal incision was made at the distal third of the clavicle, and a distal incision was created at the neck of the humerus. For embryonic shoulder samples where distinctive bone features had not formed, approximations were made to capture the entirety of the glenohumeral and acromioclavicular joints. For the cranium samples (<8 PCW),

two regions were dissected for each of the calvaria and skull base, separated at the

posterior border of the frontal bone in both cases. For older cranial samples (>8 PCW),

tissues were dissected to separate the frontal, parietal, sphenoid, ethmoid, occipital

and temporal bones where feasible. Samples were initially embedded in optimal

cutting temperature medium (OCT) and frozen at -80°C on an isopentane-dry ice

slurry. Cryosections were then cut at a thickness of 10 µm using a Leica CM1950

cryostat and placed onto SuperFrost Plus slides (VWR) for ISS or Visium CytAssist,

or used directly for single-nuclei processing.

In situ sequencing (ISS) and high-resolution imaging

ISS was performed using the 10X Genomics CARTANA HS Library Preparation Kit

(1110-02, following protocol D025) and In Situ Sequencing Kit (3110-02, following

protocol D100), which comprise a commercialised version of HybRISS[117]. Probe panel

design was based on fold-change thresholds in cell states of the limbs (Supplementary

Table 3). Briefly: cryosections of developing limbs were fixed in 3.7% formaldehyde

(Merck 252549) in PBS for 30 minutes and washed twice in PBS for 1 minute each

prior to permeabilization: sections were briefly digested with 0.5 mg/ml pepsin (Merck

P7012) in 0.1 M HCl (Fisher 10325710) at 37°C for 15 seconds (5 PCW) or 30 seconds

(6 PCW and older), then washed twice again in PBS, all at room temperature.

Following dehydration in 70% and 100% ethanol for 2 minutes each, a 9 mm diameter

(50 µl volume) SecureSeal hybridisation chamber (Grace Bio-Labs GBL621505-20EA)

was adhered to each slide and used to hold subsequent reaction mixtures. Following

rehydration in buffer WB3, probe hybridisation in buffer RM1 was conducted for 16

hours at 37°C. The 158-plex probe panel included 5 padlock probes per gene, the

sequences of which are proprietary (10X Genomics CARTANA). The section was

washed with PBS-T (PBS with 0.05% Tween-20) twice, then with buffer WB4 for 30

minutes at 37°C, and thrice again with PBS-T. Probe ligation in RM2 was conducted

for 2 hours at 37°C and the section washed thrice with PBS-T, then rolling circle

amplification in RM3 was conducted for 18 hours at 30°C. Following PBS-T washes,

all rolling circle products (RCPs) were hybridised with LM (Cy5 labelling mix with DAPI)

for 30 minutes at room temperature, the section was washed with PBS-T and

dehydrated with 70% and 100% ethanol. The hybridisation chamber was removed and

the slide mounted with SlowFade Gold Antifade Mountant (Thermo S36937). Imaging

of Cy5-labelled RCPs at this stage acted as a QC step to confirm RCP ('anchor')

generation and served to identify spots during decoding. Imaging was conducted using
a Perkin Elmer Opera Phenix Plus High-Content Screening System in confocal mode
with 1 µm z-step size, using a 63X (NA 1.15, 0.097 µm/pixel) water-immersion
objective. Channels: DAPI (excitation 375 nm, emission 435-480 nm), Atto 425 (ex.
425 nm, em. 463-501 nm), Alexa Fluor 488 (ex. 488 nm, em. 500-550 nm), Cy3 (ex.
561 nm, em. 570-630 nm), Cy5 (ex. 640 nm, em. 650-760 nm).
904
Following imaging, each slide was de-coverslipped vertically in PBS (gently, with
minimal agitation such that the coverslip 'fell' off to prevent damage to the tissue). The
section was dehydrated with 70% and 100% ethanol, and a new hybridisation chamber
secured to the slide. The previous cycle was stripped using 100% formamide (Thermo
AM9342), which was applied fresh each minute for 5 minutes, then washed with PBS-
T. Barcode labelling was conducted using two rounds of hybridisation, first an adapter
probe pool (AP mixes AP1-AP6, in subsequent cycles), then a sequencing pool (SP
mix, customised with Atto 425), each for 1 hour at 37°C with PBS-T washes in between
and after. The section was dehydrated, the chamber removed, and the slide mounted
and imaged as previously. This was repeated another five times to generate the full
dataset of 7 cycles (anchor and 6 barcode bits).
916

Multiplexed smFISH
Cryosections were processed using a Leica BOND RX to automate staining with the
RNAscope Multiplex Fluorescent Reagent Kit v2 Assay (Advanced Cell Diagnostics,
Bio-Techne), according to the manufacturers' instructions. Probes may be found in
Supplementary Table 7. Prior to staining, fresh frozen sections were post-fixed in 4%
paraformaldehyde in PBS for 45 minutes at 4°C, then dehydrated through a series of
50%, 70%, 100%, and 100% ethanol, for 5 minutes each. Following manual pre-
treatment, automated processing included digestion with Protease III for 15 minutes
prior to probe hybridisation. Tyramide signal amplification with Opal 520, Opal 570,
and Opal 650 (Akoya Biosciences) and TSA-biotin (TSA Plus Biotin Kit, Perkin Elmer)
and streptavidin-conjugated Atto 425 (Sigma Aldrich) was used to develop RNAscope
probe   channels.   Stained   sections   were   imaged   as   for   ISS   above.
929

Image-based in situ sequencing (ISS) decoding

We employed the ISS decoding pipeline outlined in Li et al [118]. This pipeline consists
of five distinct steps. Firstly, we performed image stitching using Acapella scripts
provided by Perkin Elmer, which generated two-dimensional maximum intensity
projections of all channels for each cycle. Next, we employed Microaligner[119] to
register all cycles based on DAPI signals using the default parameters. For cell
segmentation, we utilised a scalable algorithm that leverages CellPose[120] as the
segmentation method. The expected cell size is set to 70 pixels in diameter and further
expanded 10 pixels to mimic the cytoplasm. To decode the RNA molecules, we
employed the PoSTcode algorithm[121] with the following parameters: rna_spot_size=5,
prob_threshold=0.6, trackpy_percentile=90, trackpy_separation=2. Furthermore, we
assigned the decoded RNA molecules to segmented cells using STRtree and
subsequently generated AnnData objects following the approach described by Virshup
et al. [122]. Finally, only cells with more than 4 RNA molecules were kept for downstream
analysis.

Visium processing and library preparation

Visium CytAssist Spatial Gene Expression for Fresh Frozen (10x Genomics) was
performed following the manufacturer's protocol. Regions of interest (ROIs) were
selected based on the presence of microenvironments of bone-formation relevant to
the droplet data (e.g. coronal suture) and aligned to the CytAssist machine gasket
accordingly. Images were captured using a Hamamatsu S60 slide scanner at 40x
magnification prior to conducting the Visium CytAssist protocol for subsequent
alignment. Libraries were mapped with SpaceRanger (10x Genomics).

Single-nuclei isolation and library preparation

Single-nuclei were isolated from fresh frozen samples through cryosectioning followed
by mechanical dissociation as described in previous work [123]. Briefly, 10 μm sections
were homogenised in homogenization buffer (250 mM sucrose, 25 mM KCl, 5 mM
$MgCl_2$, 10 mM Tris-HCl, 1 mM dithiothreitol (DTT), 1× protease inhibitor, 0.4 U μl^{-1}
RNaseIn, 0.2 U μl^{-1} SUPERaseIn and 0.1% Triton X-100 in nuclease-free water) using
a glass Dounce tissue grinder set (Merck). Samples were dissociated with 10-20
strokes of a loose pestle 'A' followed by 10 strokes of a tight pestle 'B' when tissue

fragments remained . The resulting mixture was passed through a 50 μm cell strainer,

followed by centrifugation (500 g, 5 mins) the pellet was then resuspended in 300 μl

of storage buffer (1× PBS, 4% BSA and 0.2 U μl⁻¹ Protector RNaseIn) and passed

through a 20 μm cell strainer. Nuclei were visualised and assessed for viability under

microscopy following staining with trypan blue solution and were further processed for

10x Genomics Chromium Single Cell Multiome ATAC + Gene Expression according

to the manufacturer's protocol. Nuclei suspensions were loaded with a targeted nuclei

recovery of 10,000 droplets per reaction. For some of the nuclei samples, mixtures of

different donors were pooled within one reaction and later demultiplexed by genotype.

Quality control of cDNA and final libraries was done using Bioanalyzer High Sensitivity

DNA Analysis (Agilent). Libraries were sequenced using a NovaSeq 6000 (Illumina)

with a minimum sequencing depth of 20,000 read pairs per droplet.

Data preprocessing

Sequencing data were aligned to the human reference genome (GRCh38-2020-A-

2.0.0) using CellRanger-ARC software (v.2.0.0). The called barcodes from 10x

Multiome lanes with pooled genotypes from multiple donors were demultiplexed per

Genotype using BAM outputs through Souporcell (v2.0) [124]. Subsequently, the

Souporcell outputs were clustered by genotype for metadata assignment to each

barcode. Visium data were mapped to SpaceRanger (v. 1.1.0) using default input

settings, and low-res Cytassist images were aligned to hi-res microscopy images of

the processed slides using 10x Genomics LoupeBrowser (v 7.0) according to capture

frame marker regions. For gene expression data, SoupX [125] was applied to remove

background ambient RNA. For cellranger-arc called matrices that contained >16,000

droplets (exceeding the number expected from targeted droplet recovery) we

increased the estimated global rho value by 0.1 to account for the potential of

additional ambient RNA. Scrublet [126] was applied to estimate doublet probability and

a score of >0.3 was used as a cutoff value. Droplets were filtered for >200 genes, and

<5% mitochondrial and ribosomal reads.

For scATAC-seq, we applied ArchR[127] (v1.0.2) to process the outputs from Cellranger-

atac. Initial per-cell quality control was performed considering the number of unique

nuclear fragments, signal-to-background ratio and the fragment size distribution.

Moreover, cells with TSSenrichment score<7 and nFrags< 1000 were removed.

Doublets were discarded using the default settings. Initial clustering was performed at

resolution = 0.2 with top 40 dimensions from Iterative Latent Semantic Indexing (LSI).

Then, pseudo-bulk replicates were made for each broad cell type per region from the

initial clustering results. Peak calling (501 bp fixed-width peaks) was performed based

on pseudo-bulk coverages by macs2. Then, a cell-by-peak count matrix was obtained

and exported. We applied muon[128] (0.1.2) for normalization, LSI dimension reduction,

and clustering analysis with BBKNN[129] to correct the batch effects from anatomical

regions and donors for obtaining ATAC embedding. Gene scores based on chromatin

accessibility around gene bodies were calculated. Moreover, we applied MultiVI[130] to

get a joint embedding for snRNA-seq and scATAC-seq.

1007

Doublet detection

To interrogate doublet capture, we conducted an adapted scrublet workflow that was

previously described[131]. Briefly, per-droplet scrublet scores were first determined for

cellranger-arc count-matrices from each 10x multiome (gene-expression) lane

independently. The droplets were then overclustered through the standard scanpy

workflow using default parameters up to leiden clustering. Each individual cluster was

further clustered. A per-cluster median of scrublet scores were computed. A normal

distribution of doublet score, centred at the score median with a standard deviation

estimated from the median absolute deviation (MAD) was used to compute p-values

for each of the clusters. After false-discovery rate adjustment using benjamini-

hochberg correction, a p-value of >0.6 was deemed as a cutoff value of good quality

cells as doublets are significant outliers.

1020

Cell cluster annotation

We adopted a hierarchical clustering approach by first conducting Leiden clustering

on the global integrated scVI (hidden layers=256, latent variables=52,

dispersion='gene-batch') RNA embeddings to obtain broad clusters. To validate these

we utilised Celltypist to train a model on cell states in the embryonic limb bud [11,132,133],

and transferred labels onto our embedding for inspection. We utilise this information

in addition to canonical marker genes, to annotate broad clusters and subset sub-

lineages. For sub-lineages e.g. (Chondrocytes, Fibroblasts, Bone-related populations,

Schwann Cells, Immune cells, Endothelial cells), we further embedded these using

scVI (hidden layers=256, latent variables=52, dispersion='gene-batch') and conducted
Leiden clustering (resolution = 0.6), followed by DEG analyses (method='wilcoxon') to
obtain cluster markers. We additionally utilised the inferred spatial location of cell
states (described below) to inform annotations.
1034
Spatial mapping using Cell2location
We performed Cell2location for deconvolution of visium cytassist using annotated
snRNA-seq data from skull base and calvaira cell states as input. The donor was used
as batch variables, and the libraries were considered as covariates in the regression
model. For spatial mapping, we estimated 30 cells per voxel based on histological
data, and hyperparameter detection_alpha of 20 were applied for per-voxel
normalisation.
1042
ISS Patcher
ISS_patcher is a simple package for approximating features not experimentally
captured in low-dimensional data based on related, high-dimensional data. It was
developed as an approach to approximate expression signatures for genes missing in
ISS data using matched snRNA-Seq data as a reference in this study. First, a shared
feature space between both datasets was identified by subsetting the 155-158 genes
present in the ISS pool, followed by separate normalisation to median total cell counts,
log-transformation and z-scoring for both modalities. Then, the 15 nearest neighbours
in scRNA-Seq space were identified for each ISS cell with the Annoy python package,
and the genes absent from ISS were imputed as the average raw counts of the scRNA-
Seq neighbours.
Visium axis annotation using OrganAxis
Our Visium cranium sample was annotated with TissueTag by semi-automatic mode
to generate a one-dimensional maturation axis. Regions of the developing bone were
first manually annotated based on H&E features. Tissue regions that did not include
bone-forming niches were excluded from annotation. The annotation categories that
were stored included multiple regions of the coronal suture (level_0 to level_2
annotation), stemming from the central-most portion, an osteogenic front (level_3
annotation) harbouring histological features of osteoprogenitors and osteogenic zones
(level_4 to level_7 annotation) from the emergence of histological osteoblasts. All
annotations were saved as TissueTag output format, which logs the annotation

resolution, the pixels per μm (ppm) as well as pixel value interpretation of annotation
names (e.g. 0 = "Suture") and colours (e.g. "Osteogenic Front": "red"). To robustly and
efficiently migrate TissueTag annotations to the Visium objects, we first transferred
TissueTag annotations from pixel space to a high-resolution hexagonal grid space (15
μm spot diameter and 15 μm point-to-point centre distance with no gap between spots)
based on the median pixel value of the centre of the spot (radius/4) in the annotated
image. Next, to generate continuous annotations for Visium data we measured for
each spot in the hexagonal high-resolution grid the mean Euclidean distance to the 10
nearest points from each annotated structure in the level_0 annotation as well as the
distance from the closest point for structures in annotation_level_1. All annotations
were mapped to the Visium spots by proximity of the spot annotation grid to the nearest
corresponding spot in the Visium array.

1076

Gene regulatory network analysis
The SCENIC+ [134] pipeline was used to predict transcription factors and putative target
genes as well as regulatory genomic regions harbouring binding sites. The fragment
matrix of peaks called with macs2 and processed within ArchR [127] together with the
corresponding RNA count matrix were used as inputs. Meta-cells were created by
clustering cells into groups of around 10-15 cells based on their RNA profiles and
subsequent aggregation of counts and fragments. The pipeline was applied to subsets
of the dataset corresponding to individual lineages: First, CisTopic was applied to
identify region topics and differentially accessible regions from the fragment counts as
candidate regions for TF binding. CisTarget was then run to scan the regions for
transcription factor binding sites and GRNBoost2 [135] was used to link TFs and regions
to target genes based on co-expression/accessibility. Enriched TF motifs in the
regions linked to target genes were used to construct TF-region and TF-gene
regulons. Finally, regulon activity scores (AUC) were computed with AUCell based on
target-gene expression and target region accessibility, and regulon specificity scores
(RSS) derived from them. Networks of TFs, regions and target genes (eGRNs) were
constructed by linking individual regulons. TF-enhancer-gene links for all subsets
(osteogenesis, chondrogenesis, fibrogenesis, early joint progenitors, immune,
schwann) can be found in Supplementary Table 9.

1096

Trajectory analysis

For trajectory construction in the osteogenic subcompartment, non-cycling droplets were subsetted and X_scVI were used as projections for palantir in order to obtain multiuscale diffusion space. A neighbourhood graph was generated on the diffusion space using sc.pp.neighbors, and the first two PCs were used as initial positions to create ForceAtlas2 embeddings using sc.tl.draw_graph. The FA embeddings were exported into R, and monocle3 [136,137] was used to find a principal graph and define pseudotime. Differentially expressed genes were then computed along pseudotime using a graph-based test (morans' I) [138,139] and the principal graph in monocle3, which allows identification of genes upregulated at any point in pseudotime. The results were visualised with heatmaps using the complexHeatmap [140] and seriation [141] packages, after smoothing gene expression with smoothing splines in R (smooth.spline, df=12). Velocity analysis[142] was performed using scvelo[143] version 0.2.3. Spliced and unspliced read counts were computed with velocyto from the unprocessed data, before using scvelo.pp.moments, scvelo.tl.velocity and scvelo.tl.velocity_graph to compute velocities for the preprocessed droplets.

In silico TF perturbations

CellOracle [144] was used with the osteogenesis trajectory created with scFates [145] and the regulons predicted with SCENIC+ [134] for the same cells were imported into CellOracle as a base GRN. Cells were aggregated into meta-cells of 10-15 cells and linear models explaining TF- from target gene expression were fitted with CellOracle per cell cluster. Regulon-based TF perturbation vectors were inferred using the cell cluster-specific models to predict effects of TF overexpression and knockout. Diffusion pseudotime [146] was then computed for IM and EC lineages separately by selecting corresponding starting points. The pseudotime gradients were used to derive pseudotime-based differentiation vectors, and the pseudotime-perturbation vector cross-product was computed to obtain perturbation scores (PS). These PS indicate, whether the *in silico* perturbation of a TF is consistent with or opposes differentiation along a lineage (osteogenesis). The simulations were carried out systematically, overexpressing and knocking out all TFs in the GRN. For each TF and condition the PS were then averaged per cell cluster and summarised in a table to screen for TFs promoting or inhibiting osteogenesis.

fGWAS

Functional GWAS analysis (fGWAS) [147] was applied to identify disease relevant cell clusters as described in detail in [148] (https://github.com/natsuhiko/PHM). The model makes use of full summary statistics from GWAS studies, linking SNPs to genes and captures a general trend between gene expression and disease association of linked loci for each cell cluster. At the same time, it also corrects for linkage disequilibrium (LD) and other relevant factors. We used full GWAS summary statistics obtained from the EBI GWAS catalog, open targets and knee and hip osteoarthritis (OA) as well as total knee (TKR) and hip (THR) replacement from [149] (Supplementary Table 8).

SNP2Cell

We used a network propagation [92] approach to integrate GWAS summary statistics and cell cluster marker gene based scores for prioritising disease relevant and cell cluster specific subunits of our TF-network. First, scores per SNP were computed from downloaded summary statistics and weighted by linkage-disequilibrium. Then the scores were mapped to a gene regulatory network, here an eGRN computed with SCENIC+ for the corresponding lineage. Since the used networks contain TFs and target genes, and also regions with TF-binding sites as nodes, SNP scores were mapped both to genes and regions, representing distal regulatory elements. The scores were then propagated across the network using a random walk with restart (or personalised page-rank) process. This integrates the contribution of individual SNPs, with signals converging around relevant network nodes. The procedure was repeated with 1000 randomly permuted scores to compute permutation-test results and z-scores. Next, differential expression based marker gene scores for each cell cluster were propagated in the same way, resulting in cell-cluster specificity scores for each network node. The SNP and expression based scores were then combined per cell cluster as in [150] by using the minimum for each node. The final scores were thresholded and the resulting connected components obtained as enriched sub-networks. The method has been compiled into a tool we called SNP2CELL and is available at (https://github.com/Teichlab/snp2cell).

Mutation calling at single cell resolution

Scomatic was applied to detect somatic single-nucleotide mutations in snRNA-seq and snATAC-seq datasets. We first split alignment files into broad cell cluster-specific bam

files per donor per modality. Then a base count matrix indexed by cell cluster and
genomic coordinates was created and merged. We applied somatic variant calling for
each cell cluster based on the matrix to get more mutations for each modality with
snATAC. Then beta binomial test was employed to discount sites affected by recurrent
technical artefacts as somatic mutations. We also applied additional filters based on
Panel of Normals (PON)) to remove germline polymorphisms and recurrent artefacts.
At least 5 reads were covered to consider a callable site and for mutated sites, it is
supported by at least three reads from at least two cells from the same cell cluster as
mutations in cell cluster level. We set max_cell_types to 10000 to disable the filter cell
cluster number filters as we applied Genome Aggregation Database (gnomAD) to
remove suspicious germline mutations as a second criterion. We kept all the other
filtered mutations instead of using PASS mutations to keep more mutations for
downstream analysis. Mutations from RNA and ATAC were aggregated together. We
also applied the function 'SingleCellGenotype.py' to compute the genotype for each
cell at the variant sites with default parameters.

In the chondro-lineages and the Schwann compartments, we filtered mutations based
on the cell cluster level first. Shared mutations per donor require them to mutate in
only both chondro-lineages and the Schwann compartment but also to be callable in
at least one reference cell clusters across all regions (Immune, Endothelial, Epithelial,
Erythroid, PArchMes, Lymphatic endothelial, $PAX7^+$ Myo, $MYH3^+$ Myo, Synovial Fibro,
$PI16^+$ FibroPro, DermFibro, MyoFibro, $HIC1^+$ Mes, Tenocyte, LimbMes, FacialMes,
Meningeal Dura, Periosteal Dura, LEPR Mese, Perisuture, SutureMes1, SutureMes2,
CranialMes), which rules out the germline mutations again. Then we exact mutated
single cells based on the shared mutations. Cells with at least 2 reads that exhibited
mutations are treated as mutated cells. Finally, we aggregated all the mutated cells
across all donors and calculated the density of cells in an embedded space using
'tl.embedding_density' function in scanpy to get normalised density values.

For shared mutations between $PAX7^+$ Chondrocyte and muscle lineages, we filtered
mutations that only mutated in $PAX7^+$ chondrocyte and the whole muscle compartment
and the reference cell clusters were Immune, Endothelial, Epithelial, Erythroid, IZ,
PArchMes, Lymphatic endothelial, Synovial Fibro, PI16hiFibroPro, DermFibro,
MyoFibro, HIC1hiMes, Tenocyte, LimbMes, FacialMes, Meningeal Dura, Periosteal
Dura, LEPR Mes, Perisuture, SutureMes1, SutureMes2, CranialMes, Neural,
Schwann, and Perichondrium.

Cell-cell interactions

Ligand-receptor interactions were inferred using 'cpdb_analysis_method.call' in CellPhoneDB (v4.0.0). We only considered genes with more than 10% of cells demonstrating expression within each cell cluster considered and interactions with P >0.001 were removed. We used NicheNet (v1.1.1) to identify different interactions between endochondral ossification and intramembranous niches. We first calculate differentially expressed genes (DEGs) of osteoblastic cell clusters and tip cells between two niches using standard Seurat Wilcoxon test and min LFC per cell cluster were used to summarise the differentially expressed ligands and receptors. Then, top 1000 DEGs were applied for calculating the ligand activities. We prioritised the ligand-receptor links with default settings. Top 10 ligands and their top scoring receptors were visualised in heatmaps.

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a

Cranial Skeleton

Calvaria

Skullbase

Bone and cranial suture

Intramembranous

Neural Crest (Anterior) ←→ Mesoderm (Posterior)

Endochondral

Appendicular Skeleton

Shoulder Joint

Hip Joint

Knee Joint

1 Mesenchyme Condensation

2 Incipient Joint Formation

3 Chondrogenesis Zonation

○ Limb mesenchyme
● Interzone cells
● Articular chondrocytes
● Hypertrophic chondrocytes

b

PCW 5 6 7 8 9 10 11 12

● Calvaria
● Skullbase
● Shoulder
● Hip
● Knee

● Multiome
○ ISS
☆ Visium

Intramembranous ossification ●
Endochondral ossification ● ● ● ●

c

Co-embedding

10x Multiome

Open chromatin   RNA

TF motif
enhancer
promoter
mRNA

Differentiation

SCENIC+

TF → r1, r2, r3, r4 → g1, g2, g3

TF

r1   r

fGWAS

scATAC

UMAP2 / UMAP1

+

-log P value / chr

logOR

d

RNA   n = 394,021

Erythroid, Immune, Muscular, Endothelial, Epithelial, Schwann, Neural, Mesenchyme

ATAC   n = 483,400

Muscular, Neural, Schwann, Endothelial, Epithelial, Mesenchyme, Immune, Erythroid

RNA + ATAC   n = 336,162

Neural, Muscular, Immune, Endothelial, Epithelial, Schwann, Mesenchyme, Erythroid

UMAP1 / UMAP2

e

Skeletal region composition

1.0

0.5

0.0

Calvaria Skullbase Shoulder Hip Knee

● Mesenchyme
● Muscle
● Schwann
● Immune
● Erythroid
● Epithelial
● Neural
● Endothelial

f

Interzone formation

RUNX2
GDF5

5.7 PCW leg, ISS

Cartilage primodria formation

ACAN
SOX9

6.7 PCW arm, ISS

Perichondrium formation

ACAN
THBS2

ACAN
RUNX2

7.3 PCW knee, ISS

***Cartilage and bone maturation**

ACAN
MATN1

ACAN
IBSP

7.3 PCW arm, ISS

Figure. 1: Human fetal skeletal atlas overview. a) Anatomical sampling approach to dissect
five main anatomical sites. Cell types of origin within the cranium are determined by the
anterior-posterior axis. **b)** Donor overview across age and anatomical regions sampled;
atlasing modalities represented in the legend. **c)** Analysis approaches applied to integrated
RNA-ATAC seq data. **d)** UMAP embedding of dataset using RNA-seq only, ATAC-seq only
and integrated datasets. **e)** Cellular compartment composition of the skeletal region across
anatomical regions. **f)** Relative cell type abundance across anatomical locations.

Figure. 2: Formation of the embryonic joints across space and time. a) Differential
abundance (MILO) of cell states across lineages against multiple comparators. Top: Broad
clusters of non-myogenic mesenchyme. Bottom: Cranial osteogenic subclusters. ChondroPro:
Chondroprogenitors, *EPYC*$^+$ Chondro: *EPYC*$^+$ Chondrocyte, Articular Chondro: Chondrocyte,
Hypertrophic Chondro: Hypertrophic Chondrocyte, Maturing Chondro: Maturing Chondrocyte,
PAX7$^+$ Chondro: *PAX7*$^+$ Chondrocyte, *HIC1*$^+$ Mes: *HIC1*$^+$ Mesenchyme, FibroPro: Fibroblast
progenitors, Synovial Fibro: Synovial fibroblast, DermFibro: Dermal fibroblast, MyoFibro:
Myofibroblast, *PAX9*$^+$ Mes: *PAX9*$^+$ Mesenchyme, *ALX1*$^+$ Mes, *ALX1*$^+$ Mesenchyme,
CranialMes: Cranial mesenchyme, FacialMes: Facial mesenchyme, PArchMes: Pharyngeal
mesenchyme, *LEPR*$^+$ Mes: *LEPR*$^+$ Mesenchyme, SutureMes1: Suture mesenchyme1,
SutureMes2: Suture mesenchyme2, *HHIP*$^+$ PreOB: *HHIP*$^+$ Pre-Osteoblast. Mixed: other
related cell types that are not included in each plot respectively. **b)** Normalised gene
expression of genes in modules associated with skeletal development within InterzoneChon
subclusters. **c)** RNA-velocity on UMAP of RNA-sub-clustered cell states of the broad
InterzoneChon cluster. **d)** SCENIC+ predicted TF expression (box color), dot size shows
target gene accessibility (AUCell) and dot shade (grayscale) shows target gene expression
(GEX AUCell). **e)** Demultiplexed marker gene expression from ISS of the 5.7 PCW lower limb.
f) ISS-Patcher workflow (see methods). **g)** Spatial plots of ISS-Patcher cell cluster imputation.
h) Schematic of knee joint interzone formation, gradients demonstrating genes associated
with zonated hypertrophic and articular phenotypes. **i)** Histological view and annotations of
adjacent sections of 10x cytassist visium data. Cell2location results of coronal suture and
schematic demonstrating TF gradient across regions of the coronal suture. **j)** SCENIC+
predicted TF expression (box color), dot size shows target gene accessibility (AUCell) and dot
shade (grayscale) shows target gene expression (GEX AUCell) in suture progenitors. **k)**
Coverage plots showing aggregated scATAC signals around the *HHIP* and *RUNX2* loci for
osteoprogenitor cell states with increasing osteogenic phenotype in IM (CranialMes to HHIP+
ProOB) and EC (ArticularIZ to Perichondrium). Below each coverage plot, loops predicted by
Scenic+ between TSS and enhancers are shown (colored by importance score). Selected
upstream TFs predicted by Scenic+ to bind and regulate via some of the enhancers are shown
on the left. The network links inhibitors of osteogenesis (such as TWIST1, LMX1B) to pro-
osteogenic genes (RUNX2, HHIP) via overall inhibitory connections.

Figure 3: Intramembranous and endochondral osteogenesis. a) Force directed (FA) embedding showing trajectories per anatomical region that converge on osteoblasts and osteocytes. In intramembranous (IM) ossification the suture mesenchyme constitutes a distinct state that maintains progenitor pools. A two-stage IM process forms the sutures then subsequently bone-lineage. LimbMes: Limb mesenchyme. **b)** Heatmaps showing the expression of differentially expressed genes along pseudotime for IM and endochondral (EC) ossification. Differential expression has been tested using a spatial autocorrelation test in monocle3. **c)** TF accessibility (AUCell ATAC) and Target gene accessibility (AUCell GEX) of selected TFs along pseudotime. Differing patterns are observed for osteogenic (RUNX2) and inhibitory TFs (TWIST1, LMX1B, NFATC2), pointing to a complex regulatory mechanism. **d)** Spatial binning of Visium Cytassist image of frontal bone (sagittal section) using OrganAxis, axis begins from coronal suture to the ossifying frontal bone (posterior-anterior) Left-Right: Histological image of frontal bone; Cell-cluster enrichment from Cell2location; Axis values (rainbow); manual bins based on histological features **e)** Enrichment within spatial bins for IM cell states (left), selected marker gene expression (middle) and pathway enrichment (right). **f)** 7.3 PCW humerus (Coronal section) and imputed cell clusters showing EC ossification as a sequence of mapped cell-states. ACr, Acromion; Cor, Coracoid; H, Humerus; Sk, Skin. **g)** Cross-section of Visium Cytassist at the meeting point between calvaria and skull base (anterior border of the sphenoid), chondrogenic cell states enrichment from Cell2location **h)** Predicted ligand-receptor interactions per IM and EC cell clusters. **i)** Gene expression along IM and EC cell clusters in a) as well as endothelial cells. **j)** Ligand-receptor interactions predicted between endothelial cells and cells of the osteogenic lineage.

Figure. 4: Chondrogenesis across different anatomical regions. **a**) RNA UMAP embeddings of chondrocytes coloured by subtypes. **b**) Dot plot showing the mean expression (dot colour) and fraction of expressing cells (dot size) of selected marker genes. **c**) Distribution of different chondrogenic sub types across anatomical regions. a, b, c share the same legends. **d**) RNA UMAPs (the upper panel) of representative chondrogenic genes (*ACAN*, *COL2A1*) and myogenic (*PAX7*, *MEGF10*) for *PAX7*$^+$ chondrocytes. Genome browser tracks (the lower panel) from scATAC-seq analyses displaying the sum signal within the indicated gene loci. OtherChon represents the whole chondro-lineages except *PAX7*$^+$ Chon. **e**) RNA scope for cells that co-expressed myogenic markers (*PAX7*, red) and chondrogenic markers (ACAN, green) in the shoulder at 7.3PCW. The blue marks the cell nucleus. Scale bar, 20/200 μm. **f**) Enhancer-GRN showing enriched TFs in *PAX7*$^+$ Chon, inferred using SCENIC+. Circles and diamonds represent genes and regions with TF-binding sites, respectively. Colour indicates whether the TF of *PAX7*$^+$ Chon is also differentially expressed in chondro-lineages or *PAX7*$^+$ myocytes. TF-region links are coloured by TF. **g**) The left panel shows the UMAP of myogenic lineages, while the right panel shows the gaussian kernel density estimates of shared somatic mutations over myogenic lineages and chondrocyte lineages. Myoprog: Myogenic progenitor. **h**) Barplots showing the number of shared mutations in myogenic cells per anatomical site/pcw/donor/celltype. **i**). Schematic representation of the putative non-classical lineages that contribute to chondrogenesis.

Figure 5: Links to complex diseases of the skeleton. a) fGWAS analysis results for
enrichment of GWAS signals for bone-related complex traits in broadly annotated cell states.
b) Enrichment of selected complex traits in finely annotated osteo- and chondrogenic cell
states (top) and InterzoneChon subclusters (bottom). **c)** Schematic representation of
SNP2cell: scores derived from GWAS summary statistics and cell cluster marker scores are
mapped and integrated across a gene regulatory network, highlighting enriched modules
predicted to play a cluster-specific role in disease. **d)** Enrichment scores for knee OA across
all cell clusters (heatmap, top) showing a cluster of similar enriched genes for chondrocytes
and median enrichment scores (boxplot, bottom). **e)** Enrichment scores for hip OA across the
osteogenic trajectory (heatmap, top) and median enrichment scores (boxplot, bottom). **f)**
Articular chondrocyte and knee OA specific enriched sub-network. Brighter colours
correspond to a larger enrichment score, relative to scores obtained from random
permutations. **g)** Perichondrium and hip OA specific enriched sub-network. Brighter colours
correspond to larger enrichment scores as in d. **h)** Gene Set Enrichment Analysis for GO
Biological Process terms across Articular Chondro-knee OA and Perichondrium-hip OA
enrichment scores.

Figure 6: Links to Monogenic conditions affecting bone development. a) Known genes mutated in musculoskeletal monogenic conditions with their expression and accessibility along the intramembranous and endochondral osteogenesis trajectory. **b)** Force-directed embedding of the endochondral trajectory showing cell cluster annotations. **c)** Heatmap of *in silico* TF knockout perturbation scores per cell cluster. A higher score indicates transcriptomic changes induced by the perturbation are promoting osteogenesis. **d)** TF perturbation vectors showing the direction of induced transcriptomic changes on the FA embedding for three selected TFs: TWIST1, MSX2 and LMX1B. KO simulation of all three genes promotes osteogenesis, while overexpression inhibits it. **e)** TF expression on Visium Cytassist of the frontal region of skull, enrichment of TFs in coronal suture and skull-forming layers. **f)** Expression of synostosis-associated genes along spatial zones defined using OrganAxis. **g)** Enhancer-GRN showing inter-regulation between TWIST1, MSX2 and LMX1B, inferred using SCENIC+. Circles and diamonds represent genes and regions with TF-binding sites, respectively. Region-gene links are coloured and scaled according to peak2gene importance, while TF-region links are coloured by TF. Blue circles represent regulated targets, orange circles regulators and white intermediate genes. **h)** Enhancer-GRN showing predicted regulation of SOST. Region-gene links are coloured and scaled according to peak2gene importance. Highlighted in red is a region containing the enhancer known to be mutated in van Buchem disease.