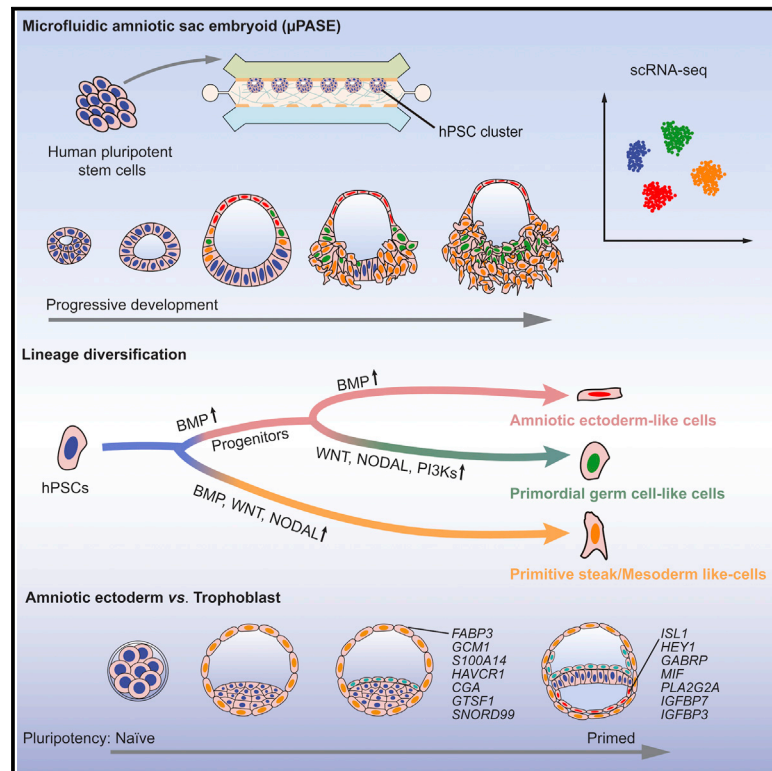# Single-cell analysis of embryoids reveals lineage diversification roadmaps of early human development

Graphical abstract

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In brief

Fu and colleagues profiled a stem cell-derived microfluidic amniotic sac embryoid (µPASE) using single-cell RNA sequencing. Molecular maps of lineage diversifications from the pluripotent human epiblast toward the amniotic ectoderm, primitive streak/mesoderm, and primordial germ cells were constructed and compared with *in vivo* primate data.

Highlights

- Single-cell transcriptional atlas of microfluidic amniotic sac embryoid (µPASE)

- Single-cell transcriptional characterization of chimpanzee µPASE

- Critical role of NODAL in human mesoderm and primordial germ cell specification

- Stringent criteria for distinguishing between trophectoderm and amniotic ectoderm

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Single-cell analysis of embryoids reveals lineage diversification roadmaps of early human development

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SUMMARY

Despite its clinical and fundamental importance, our understanding of early human development remains limited. Stem cell-derived, embryo-like structures (or embryoids) allowing studies of early development without using natural embryos can potentially help fill the knowledge gap of human development. Herein, transcriptome at the single-cell level of a human embryoid model was profiled at different time points. Molecular maps of lineage diversifications from the pluripotent human epiblast toward the amniotic ectoderm, primitive streak/mesoderm, and primordial germ cells were constructed and compared with *in vivo* primate data. The comparative transcriptome analyses reveal a critical role of NODAL signaling in human mesoderm and primordial germ cell specification, which is further functionally validated. Through comparative transcriptome analyses and validations with human blastocysts and *in vitro* cultured cynomolgus embryos, we further proposed stringent criteria for distinguishing between human blastocyst trophectoderm and early amniotic ectoderm cells.

INTRODUCTION

Development of multicellular organisms is one of nature's greatest triumphs. Development is a tightly orchestrated process, following stereotypic lineage diversifications and morphogenetic tissue patterning events in a precise spatiotemporal order. Scientists commonly use animal models to study the key transcriptional and signaling activities that underlie pattern formation, morphogenesis, cell differentiation, and tissue growth (Gilbert, 2000; Schoenwolf et al., 2020; Solnica-Krezel, 2020). However, cross-species genetic and morphological divergence is evident between humans and commonly used animal models (Rossant, 2015; Rossant and Tam, 2017). To address this issue, there is a significant current interest in improving *in vitro* culture protocols of human and non-human primate (NHP) monkey embryos for experimental observations and mechanistic studies (Deglincerti et al., 2016; Ma et al., 2019; Niu et al., 2019; Shahbazi et al.,

2016; Xiang et al., 2020; Yang et al., 2021). However, experimentation on human and NHP monkey embryos remains challenging because of limited access to and bioethical constraints on these natural specimens (Clark et al., 2021; Hyun et al., 2016; Lovell-Badge et al., 2021). As such, knowledge of human development remains limited; this is particularly true for early post-implantation human development, when the basic human body plan is laid down and when the human embryo *in vivo* is at its most inaccessible phase for experimentation.

Recently, stem cell-derived embryo-like structures (or embryoids) that could recapitulate certain aspects of mammalian early embryogenesis are emerging as tractable experimental tools for studying human development (Beccari et al., 2018; Haremaki et al., 2019; Harrison et al., 2017; Liu et al., 2021; Moris et al., 2020; Rivron et al., 2018; Shao et al., 2017a, 2017b; Simunovic et al., 2019; Warmflash et al., 2014; Xue et al., 2018; Yanagida et al., 2021; Yu et al., 2021; Zheng et al., 2019a, 2019b).
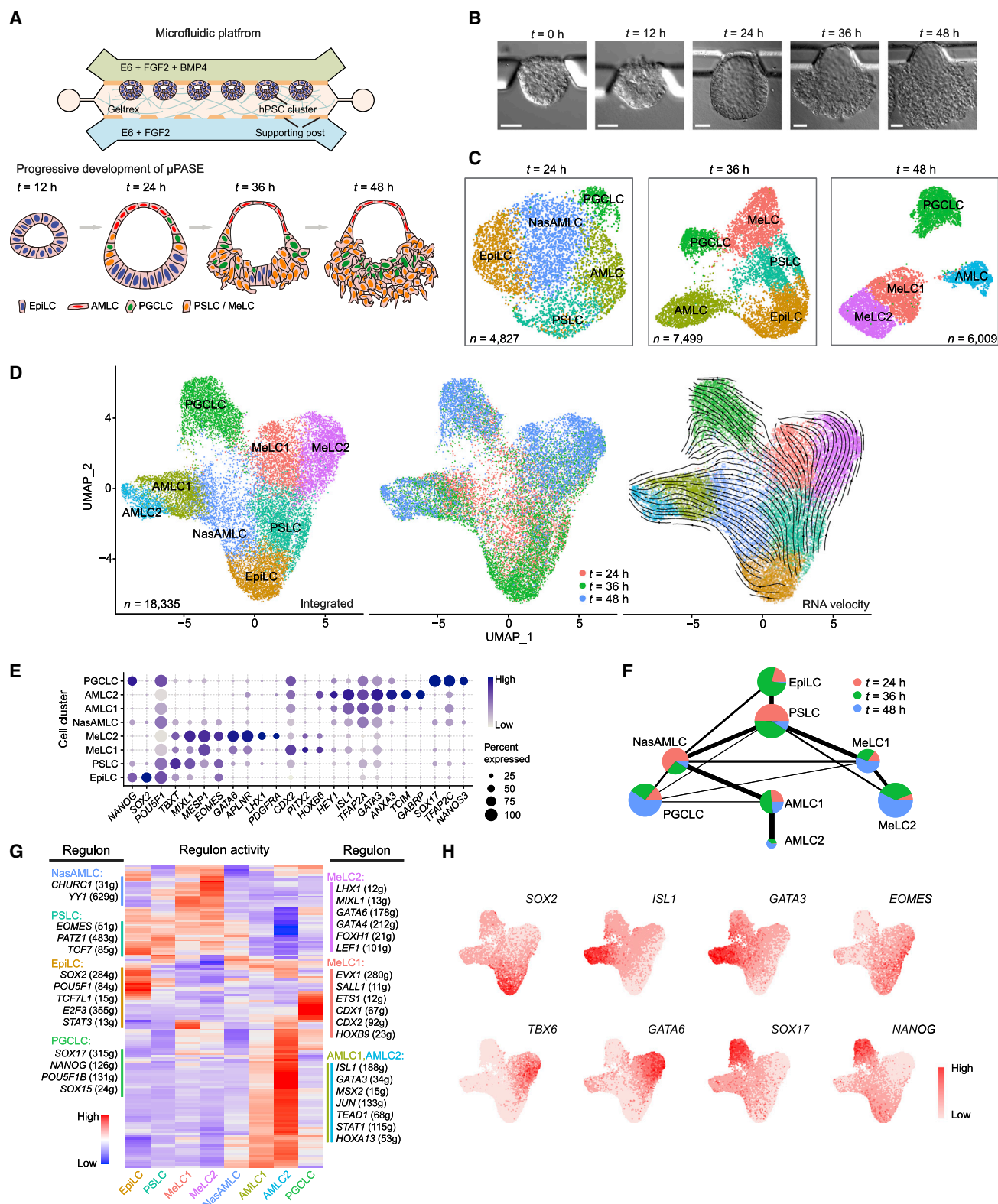
Figure 1. Single-cell transcriptomic profiling of μPASE development

(A) Development of μPASEs. Single hPSCs were guided to form uniform-sized clusters at prescribed locations in the microfluidic device. Asymmetric stimulation with exogenous BMP4 from $t = 0$–48 h led to progressive development of μPASEs. E6, Essential 6. See STAR Methods for the μPASE protocol.

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Particularly, we have recently developed a microfluidic post-implantation amniotic sac embryoid (μPASE) system (Zheng et al., 2019b), allowing the development of PASEs in a highly controllable, reproducible, and scalable fashion.

RESULTS

Single-cell transcriptomic profiling of μPASE development

Development of a μPASE involves successive cell morphogenetic and lineage specification events that recapitulate early post-implantation human development up to the onset of gastrulation (Figure 1A) (Zheng et al., 2019b, 2021). Specifically, because of their intrinsic lumenogenic property, each human pluripotent stem cell (hPSC) cluster undergoes lumenogenesis and epithelization to establish apical-basal polarity and form a single central apical lumen by t = 12 h (Figures 1B and S1A). hPSCs exposed directly to exogenous BMP4 stimulation in each cluster initiate amniogenesis, evidenced by continuous flattening of cell morphology, resolving into a thin layer of squamous amniotic cells (Figure 1B). Inductive effects of amniotic ectoderm-like cells (AMLCs) in the μPASEs lead hPSCs at the opposite pole to undergo epithelial-mesenchymal transition (EMT) and gastrulation-like events (Zheng et al., 2019b, 2021), with gastrulating cells disseminating away from μPASEs from t = 36 h onwards (Figure 1B), leading to disintegration of the μPASE structure. By t = 48 h, the μPASE contains only AMLCs, mesoderm-like cells (MeLCs), and primordial germ cell-like cells (PGCLCs), without the presence of epiblast-like cells (EpiLCs) (Figures 1B and 1C).

To investigate dynamics of μPASE development at the transcriptome level, single-cell suspensions of μPASEs at t = 24, 36, and 48 h were prepared before single-cell RNA-sequencing (scRNA-seq) using 10× Genomics platform. We performed uniform manifold approximation and projection (UMAP) dimension reduction using the Seurat R package (Butler et al., 2018; Satija et al., 2015) for scRNA-seq datasets at each time point (Figure 1C) as well as for the integrated scRNA-seq dataset from all three time points (Figure 1D). These analyses reveal distinct cell clusters in the μPASE based on expression patterns of key lineage markers (Figures 1C–1E, S1B, and S1C). Consistent with our previous findings (Zheng et al., 2019b, 2021), in the μPASE, hPSCs develop progressively from a pluripotent EpiLC stage to three distinct cell populations by t = 48 h: AMLCs, MeLCs, and PGCLCs (Figures 1C–1E, S1D, and S1E).

To understand how hPSCs transit from the EpiLC stage to AMLCs, MeLCs, and PGCLCs during μPASE development, RNA velocity analysis (Bergen et al., 2020; La Manno et al., 2018) was conducted for the integrated scRNA-seq dataset, with RNA velocity vectors overlaid on the integrated UMAP plot (Figure 1D). This RNA velocity analysis reveals developmental trajectories of the AMLC lineage (EpiLC → nascent AMLC or NasAMLC → AMLC1 → AMLC2) and MeLC lineage (EpiLC → primitive streak-like cell or PSLC → MeLC1/MeLC2) (Figure 1D). However, developmental trajectory of PGCLCs is not as clear from the RNA velocity analysis (Figure 1D). To further examine lineage relations between different μPASE cell clusters, partition-based graph abstraction (PAGA) analysis (Wolf et al., 2018) was conducted, revealing that PGCLCs correlate best with the NasAMLC cluster (Figure 1F). To reveal gene regulatory networks (GRNs) underlying each cell cluster, we performed GRN analysis using single-cell regulatory network inference and clustering (SCENIC; Aibar et al., 2017; Figures 1G and 1H).

Trajectory inference and gene expression dynamics analysis

To infer developmental trajectories of different μPASE cell lineages, we plotted a three-dimensional (3D) diffusion map based on principal-component analysis (PCA) embeddings of the integrated scRNA-seq dataset (Figure 2A and Video S1) (Angerer et al., 2016). The AMLC, MeLC, and PGCLC lineages display distinct and well-separated trajectories in the 3D diffusion map (Figure 2A). To analyze transcriptome dynamics during AMLC lineage development, EpiLC, NasAMLC, and AMLC1/2 clusters were isolated from the integrated scRNA-seq dataset and re-plotted using two-dimensional (2D) diffusion maps (Figure 2B). Similarly, to analyze MeLC lineage development, EpiLC, PSLC, and MeLC1/2 clusters were isolated from the integrated scRNA-seq dataset before re-plotting using 2D diffusion maps (Figure 2C). Expression dynamics of selected genes related to amniotic/embryonic ectoderm (AM) and PS or Meso were plotted against diffusion pseudotime (dpt) (Figures 2D and 2E, S2A, and S2B). Notably, in the AMLC lineage, expression of *TFAP2A*, *MSX2*, and *ID2*, which are commonly used AM markers (Ma et al.,

(B) Bright-field images showing progressive development of μPASEs over time, including thinning and flattening of the incipient amniotic ectoderm and thickening of the incipient mesoderm cells before their dissemination from the μPASE structure.

(C) Dimension reduction presentation via UMAP and cell identity annotations of single-cell transcriptome datasets obtained for μPASEs at indicated time points. *n* indicates cell numbers analyzed for each time point.

(D) UMAP of integrated single-cell transcriptome datasets of μPASEs from t = 24, 36, and 48 h (shown in (C)), color-coded according to cell identity annotations (*left*) or time points (*middle*). RNA velocity vectors were projected onto the UMAP-based embeddings (*right*). *n* indicates the total number of cells combined from the three time points.

(E) Dot plot showing expression of key marker genes across the cell clusters as indicated. The sizes and colors of dots indicate the proportion of cells expressing the corresponding genes and their averaged scaled values of log-transformed expression, respectively.

(F) PAGA analysis of the integrated single-cell transcriptome dataset shown in (D). The thickness of lines connecting pairs of cell clusters indicates the degree of correlation between the cell cluster pairs. Lines with a correlation weight less than 0.05 are not shown. Pie charts for each cell cluster show percentages of indicated cell types from the three time points. Pie chart size is proportional to the total number of indicated cell types. See Mendeley Data Table 1.

(G) Heatmap of regulon activities calculated from gene regulatory network interference. Selected master regulators of different cell clusters are depicted as indicated. See Mendeley Data Table 2 and Document S1.

(H) Gene set activity of selected regulons overlaid on the integrated UMAP plot from (D).

EpiLC, epiblast-like cell; PSLC, primitive streak-like cell; MeLC1/2, mesoderm-like cell 1/2; AMLC1/2, amniotic ectoderm-like cell 1/2; NasAMLC, nascent amniotic ectoderm-like cell; PGCLC, primordial germ cell-like cell. In (B), experiments were repeated more than twenty times with similar results. Scale bars, 50 μm.
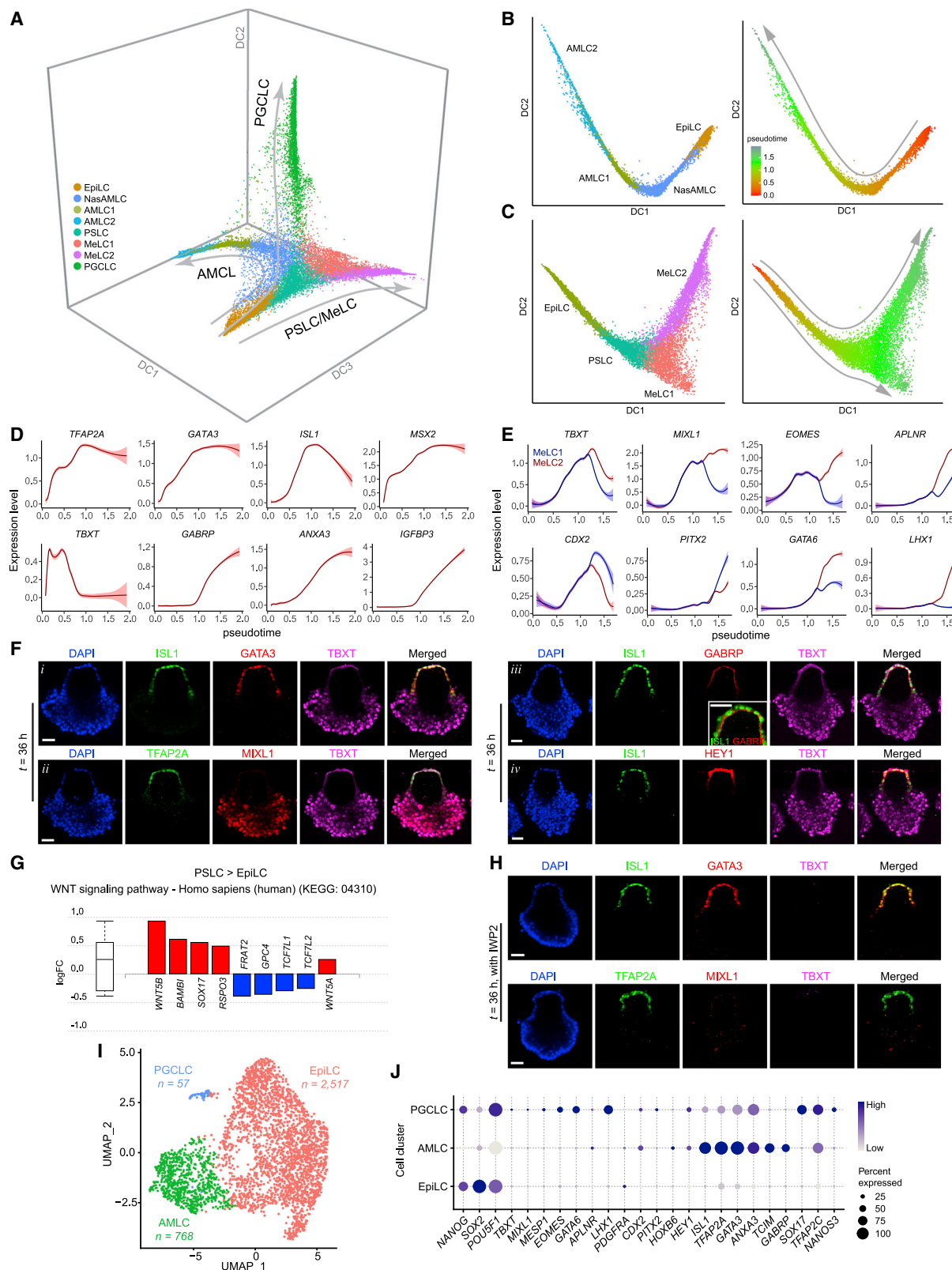
Figure 2. Trajectory inference and gene expression dynamics during μPASE development

(A) Three-dimensional diffusion map based on embeddings of the UMAP plot in Figure 1D, showing developmental trajectories of AMLC, PSLC/MeLC, and PGCLC lineages. The UMAP plot is color-coded according to cell identity annotations. See Video S1.

(legend continued on next page)

2019; Sasaki et al., 2016; Yang et al., 2021), is quickly upregulated following exogenous BMP4 stimulation (Figures 2D and S2A). Similarly, *GATA3* and *ISL1* become upregulated relatively early during AMLC lineage development (Figures 2D and S2A). In contrast, expression of *GABRP*, *IGFBP3,* and *WNT6* shows delayed upregulation, whereas *TBXT* is only transiently expressed during early AMLC lineage development (Figures 2D and S2A).

On the 2D diffusion map, the MeLC lineage branches into two separate paths corresponding to MeLC1 and MeLC2 clusters (Figure 2C). Both MeLC1 and MeLC2 lineages show upregulated expression of common Meso markers, yet some key genes exhibit distinct expression levels (Figures 2E and S2B). For example, compared with MeLC1, MeLC2 expresses relatively higher levels of *MIXL1*, *EOMES,* and *GATA6* (Figures 2E and S2B), presumably corresponding to a lateral plate or intermediate Meso fate (Daoud et al., 2014; Prummel et al., 2019). MeLC1 lineage, in contrast, expresses higher levels of *CDX2* and *HOXB6* (Figures 2E and S2B), presumably corresponding to a paraxial Meso fate (Casaca et al., 2016; Chawengsaksophak et al., 2004). We further conducted immunostaining for selected AM and Meso markers in µPASEs at t = 24, 36, and 48 h to validate lineage fate specification (Figures 2F and S2C–S2E). GATA6^{high}TBXT^{high}MIXL1^{high}CDX2^{low} MeLC2 appears as leading cells in the migratory gastrulating cell population, whereas GATA6^{low}TBXT^{low}MIXL1^{low}CDX2^{high} MeLC1 appears as trailing cells in this population (Figure S2F).

Patterning of AMLCs and PSLCs was evident in the µPASE by t = 24 h, with positive immunostaining for ISL1, GATA3, and TFAP2A in incipient AMLCs and for TBXT and MIXL1 in incipient MeLCs (Figures S2C and S2D; Video S2). Notably, expression of AM marker GABRP (Yang et al., 2021) is restricted on the apical surface of AMLCs, whereas expression of HEY1, another AM marker (Yang et al., 2021), is evident in both the nucleus and cytoplasm of AMLCs (Figure 2F). AMLCs appear to actively proliferate during µPASE development (Figure S2G). Nonetheless, we could not exclude the possibility that EpiLCs continuously supply AMLCs during µPASE development. To reveal transcriptome changes during AMLC development, we conducted differentially expressed gene (DEG) and pathway enrichment analysis to compare NasAMLC, AMLC1, and AMLC2 (Figure S2H and Mendeley Data Table 3).

To explore potential mechanisms underlying lineage choices between NasAMLCs and PSLCs for EpiLCs, we examined DEGs upregulated in NasAMLCs and PSLCs relative to EpiLCs and noticed significant overlaps (144 out of 342 genes for NasAMLCs, 144 out of 184 genes for PSLCs) (Figure S2I and Mendeley Data Table 3). KEGG pathway analysis suggests that development of both NasAMLCs and PSLCs requires transforming growth factor β signaling (KEGG: 04,350; Figure S2J), which is not surprising given that exogenous BMP4 is supplemented in the µPASE protocol. However, WNT signaling pathway (KEGG: 04,310) is evident only in PSLCs, not in NasAMLCs (Figure 2G), suggesting a critical role of WNT signaling in Meso induction but not in amniogenesis. When IWP2, a small molecule inhibitor blocking the transport, secretion, or signaling activity of all WNT molecules (Chen et al., 2009; Kadowaki et al., 1996; Zhai et al., 2004), was supplemented into the microfluidic device, development of PSLCs and MeLCs in the µPASE was completely inhibited, as evidenced by negative immunostaining for TBXT or MIXL1 (Figure 2H). In contrast, AMLCs still emerge under IWP2 treatment, as evidenced by positive immunostaining for ISL1, TFAP2A, and GATA3 in flattened, presumptive AMLCs directly exposed to exogenous BMP4 stimulation (Figure 2H). Consistently, the scRNA-seq data obtained from IWP2-treated µPASEs at t = 48 h show absence of PSLCs and MeLCs; instead, the majority of cells remain pluripotent, retaining the EpiLC identity (Figures 2I and 2J), further supporting the requirement of WNT signaling in PSLC/MeLC development. It should be noted that under IWP2 treatment, AMLCs show lower expression of several AM marker genes, such as *ISL1, GABRP,* and *GATA3*, and there are much fewer PGCLCs in IWP2-treated µPASEs (Figures S2K and 2I and Mendeley Data Table 3).

PGCLC specification

During µPASE development, incipient PGCLCs emerge together with NasAMLCs and PSLCs by t = 24 h (Figure 1C). To explore the origin and specification of PGCLCs in the µPASE, a 2D diffusion map with EpiLC, PSLC, MeLC1, and PGCLC clusters isolated from the integrated scRNA-seq dataset was plotted, which, however, did not show a continuous developmental trajectory connecting EpiLCs, PSLCs, or MeLCs with PGCLCs (Figure S3A). In addition, in this 2D diffusion map, a K-branch algorithm (Chlis et al., 2017) did not identify a branching point or branches with proper confidence (Figure S3A). Thus, PGCLCs in the µPASE are unlikely to originate from PSLCs or MeLCs.

(B) Left: trajectory inference (diffusion map) of AMLC lineage (EpiLC, NasAMLC, AMLC1, and AMLC2). Right: pseudotime analysis (color-coded) based on the AMLC lineage diffusion map.

(C) Left: trajectory inferences (diffusion map) of PSLC/MeLC lineage (EpiLC, PSLC, and MeLC1 or MeLC2). Right: pseudotime analysis (color-coded) based on the PSLC/MeLC lineage diffusion map.

(D) Expression dynamics (pseudotime) of selected genes during AMLC lineage development. Level of confidence (0.95) is indicated by band width.

(E) Expression dynamics (pseudotime) of selected genes during PSLC/MeLC lineage development. Level of confidence (0.95) is indicated by band width.

(F) Representative confocal micrographs showing µPASEs at t = 36 h stained for ISL1, GATA3, and TBXT (i); TFAP2A, MIXL1, and TBXT (ii); ISL1, GABRP, and TBXT (iii, with zoom-in view showing apical expression of GABRP); ISL1, HEY1, and TBXT (iv).

(G) Differentially expressed genes (DEGs) related to WNT signaling pathway (KEGG: 04310) in PSLC compared to EpiLC.

(H) Representative confocal micrographs showing µPASEs at t = 36 h stained for ISL1, GATA3, and TBXT (*top*) or TFAP2A, MIXL1, and TBXT (*bottom*) with IWP2 supplemented into the basal medium from t = 0 h.

(I) UMAP and cell identity annotations of single-cell transcriptome data obtained for µPASEs at t = 48 h with IWP2 supplemented into the basal medium from t = 0 h. n indicates cell numbers of indicated cell types.

(J) Dot plot showing expression of key marker genes across the cell clusters as indicated. The sizes and colors of dots indicate the proportion of cells expressing the corresponding genes and their averaged scaled values of log-transformed expression, respectively.

In (F) and (H), experiments were repeated four times with similar results. Nuclei were counterstained with DAPI. Scale bars, 50 µm.

In contrast, 2D diffusion map analysis with EpiLC, NasAMLC, AMLC1/2, and PGCLC clusters clearly shows lineage progression and bifurcation of NasAMLCs into PGCLCs and AMLCs (Figure 3A). RNA velocity vectors were also computed and overlaid onto the diffusion map in Figure 3A, revealing that NasAMLCs have the closest lineage relation with PGCLCs, consistent with the PAGA analysis in Figure 1F. Our observation here is consistent with recent studies of cynomolgus monkey embryos reporting that primate PGCs could emerge in the nascent AM prior to the gastrulation (Sasaki et al., 2016). Expression of selected genes relevant to PGC development was also plotted against diffusion pseudotime, revealing their upregulated expression during PGCLC lineage development (Figures 3B and S3B). Immunofluorescence analyses of μPASEs at $t = 24$, 36, and 48 h further confirmed spatiotemporal expression of some key PGC markers, including TFAP2C, NANOG, SOX17, and BLIMP1, in incipient PGCLCs (Figures 3C and S3C). Correlation coefficient analysis based on PGC ontogenic genes identified from the cynomolgus embryo transcriptome data suggests that the transcriptome of PGCLCs in the μPASE is similar to that of day 2 hPGCLCs derived from conventional protocols (Chen et al., 2019; Sasaki et al., 2015) (Figure S3D). It is worth noting that the μPASE is essentially a posteriorized embryonic-like structure, mimicking the posterior portion of the amnion and epiblast compartments. As such, the development of μPASE gives rise to a greater number of hPGCLCs but lacks the development of ectoderm.

Using a K-branch algorithm, we identified developmental branches and the branching point for the 2D diffusion map with EpiLC, NasAMLC, AMLC1/2, and PGCLC clusters (Figure 3A). The K-branch analysis further allowed us to separate NasAMLCs into three sub-clusters, with each sub-cluster merged with EpiLCs, AMLCs, and PGCLCs, respectively, and annotated as EpiLC-branch NasAMLC, AMLC-branch NasAMLC, and PGCLC-branch NasAMLC, respectively (Figure 3A). To explore mechanisms underlying lineage choices between AMLCs and PGCLCs for NasAMLCs, DEGs upregulated in AMLC-branch NasAMLCs and PGCLC-branch NasAMLCs, as compared to EpiLC-branch NasAMLCs, were examined (Figure 3D; Mendeley Data Table 4). KEGG pathway analysis of these DEGs reveals that PGCLC-branch NasAMLCs show upregulated WNT signaling, whereas AMLC-branch NasAMLCs exhibit upregulated Hippo activity (Figure 3E). We further performed pharmacological inhibition assays to explore the roles of different signaling pathways during μPASE development. When IWP2 was supplemented into the microfluidic device, development of PGCLCs was almost completely abolished in the μPASE (Figure 3F). PGCLCs show upregulated HIF-1 signaling activity compared to PGCLC-branch NasAMLCs (Figure S3E). When HIF-1 signaling was repressed by supplementing LY294002, which functions through binding to phosphoinositide 3-kinases (PI3Ks), in the microfluidic device, the percentage of PGCLCs was significantly reduced (Figures S3F and S3G). It should be noted that even though PI3K signaling functions downstream of HIF-1 signaling, it is also involved in other intracellular pathways. Thus, the involvements of HIF-1 and PI3K signaling in hPGCLC specification warrant further investigation.

Our scRNA-seq data analysis suggests that progenitors of PGCLCs likely pass through a transient transcriptome state

similar to that of NasAMLCs before their full commitment to the PGCLC fate. Our data, however, do not exclude the possibility that cells at the junction between the AMLC and PSLC/MeLC compartments, or from the PSLC/MeLC pole of the μPASE, with a transient transcriptome state similar to that of NasAMLCs, can also give rise to PGCLCs. After all, NasAMLCs and PSLCs share a similar transcriptome, and cells in the early post-implantation human embryo might remain plastic (Chen et al., 2019; Irie et al., 2015; Kobayashi et al., 2017; Sasaki et al., 2015).

Transcriptomic comparison between μPASEs and Carnegie stage 7 human gastrula

The μPASE recapitulates certain aspects of human development from early implantation to the onset of gastrulation. Notably, single-cell transcriptome data of a Carnegie stage 7 (CS7) human gastrula recently became available (Tyser et al., 2021). Thus, we conducted transcriptomic comparison between μPASEs and the CS7 human gastrula. We first downsampled the μPASE dataset by randomly selecting 100 cells from each cluster and integrated this downsampled dataset with the CS7 human gastrula dataset (Figures S4A and S4B). Based on the transcriptome proximity in the UMAP plot, the μPASE contains cells corresponding to "Epiblast," "Primitive Streak," "Nascent Mesoderm," "Emergent Mesoderm," "Amniotic/embryonic ectoderm," and "PGC" in the human gastrula. We next selected only these cells from the CS7 human gastrula dataset to integrate with the whole μPASE dataset. A UMAP plot of the integrated scRNA-seq dataset shows cell clustering similar to that from the μPASE scRNA-seq data alone (Figure 4A). Cells from the CS7 human gastrula are evident in all cell clusters of the UMAP plot (Figure 4A). Furthermore, for each annotated cell cluster, expression patterns of key lineage markers are consistent between cells from the CS7 human gastrula and μPASEs (Figure 4B). Most of the cells from the CS7 human gastrula fall into different cell clusters consistent with their lineage annotations in the original publication (Tyser et al., 2021), except for some cells originally classified as "Epiblast" or "Primitive Streak" that fall into NasAM or AM1 clusters (Figure 4C). Notably, PGCs in the CS7 human gastrula can be easily distinguished in the PGC cluster of the integrated dataset in the UMAP plot (Figure 4A).

In the original publication of the CS7 human gastrula (Tyser et al., 2021), a UMAP plot was generated with all of the cells in the human gastrula, including those at relatively late developmental stages, such as "Hemogenic Endothelial Progenitors" and "Erythrocytes," which could negatively affect the resolution of cell clustering analysis, especially for those closely related cell lineages. To address this issue, we re-generated a UMAP plot using only the "Epiblast," "Primitive Streak," "Nascent Mesoderm," "Emergent Mesoderm," "Amniotic/embryonic ectoderm," and "PGC" clusters from the human gastrula dataset (Figure 4D). Interestingly, this UMAP plot reveals distinct cell clusters corresponding to Meso and AM, supported by feature plots showing expression patterns of key Meso and AM markers (Figures 4D and 4E). Notably, human PGCs identified in the original publication are clustered together with AM cells in the UMAP plot (Figure 4D), supporting their close lineage relation. We further isolated the Epi, AM, and PGCs from the CS7 human gastrula to generate a diffusion map (Figure 4F). Surprisingly, AM
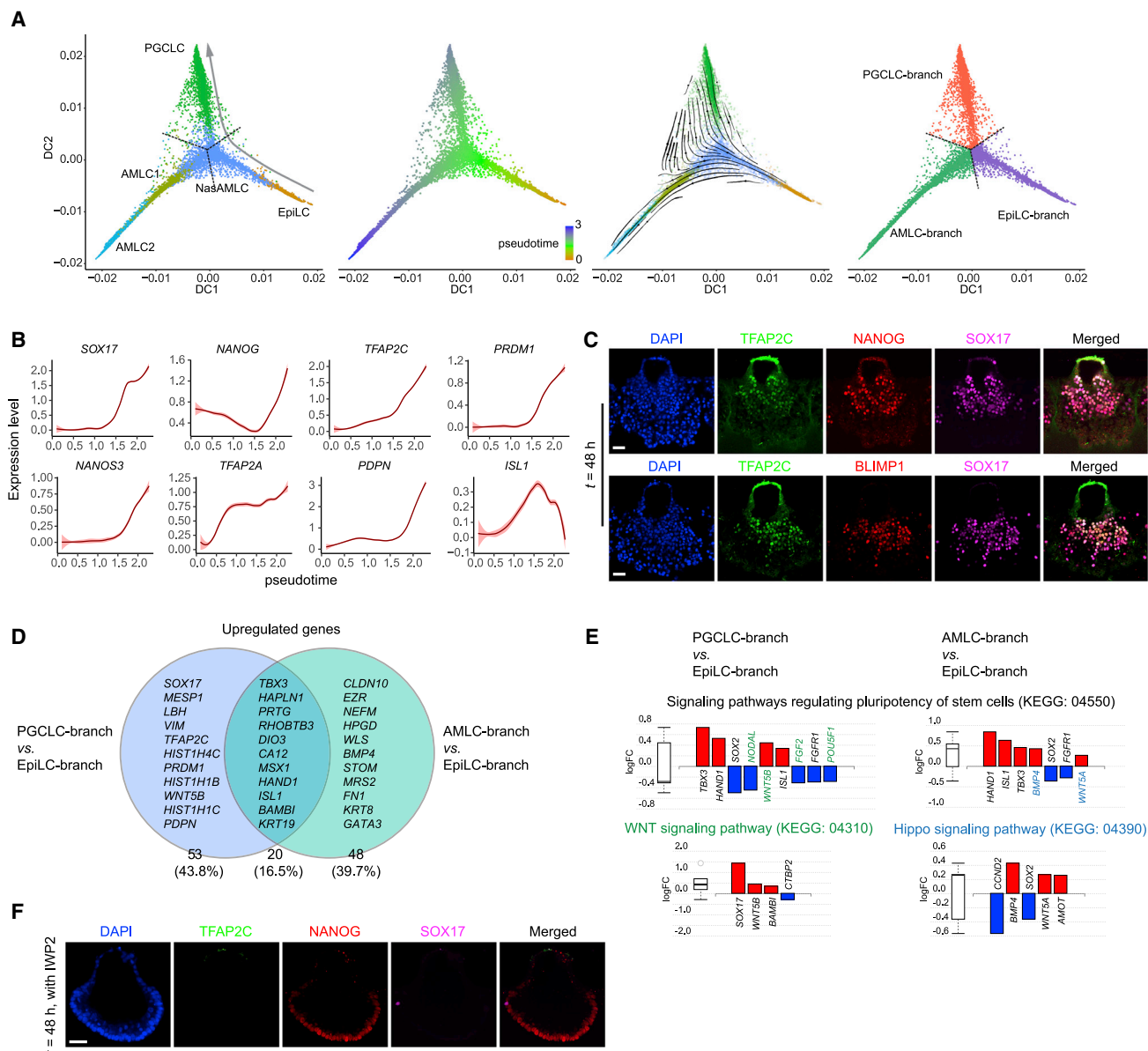
A

B

C

D

Upregulated genes

E

F

Figure 3. PGCLC specification during μPASE development

(A) Trajectory inference of PGCLC lineage. *Left:* diffusion map using EpiLC, NasAMLC, AMLC1, AMLC2, and PGCLC clusters from the UMAP plot in Figure 1D. Dotted lines show the branching point and branches identified by K-branch algorithm. *Middle left*: pseudotime analysis based on the diffusion map. *Middle right:* RNA velocity vectors overlaid on the diffusion map. *Right:* branches and the branching point identified by K-branch algorithm. Note that NasAMLC cluster is separated into three branches, which, after merging with EpiLC, AMLC1/2, and PGCLC, respectively, are annotated as EpiLC-branch NasAMLC, AMLC-branch NasAMLC, and PGCLC-branch NasAMLC, respectively.

(B) Expression dynamics (pseudotime) of selected genes during PGCLC lineage development. Level of confidence (0.95) is indicated by band width.

(C) Representative confocal micrographs showing μPASEs at t = 48 h stained for TFAP2C, NANOG, and SOX17 (top) and TFAP2C, BLIMP1, and SOX17 (bottom).

(D) Venn diagram showing upregulated genes of AMLC-branch and PGCLC-branch NasAMLCs when compared to EpiLC-branch NasAMLCs. Note that only a subset of these genes is specified in the diagram.

(E) Pathway analysis of DEGs in PGCLC-branch NasAMLCs and AMLC-branch NasAMLCs, as compared to EpiLC-branch NasAMLCs. Green and blue colors indicate genes or pathways identified only for PGCLC-branch NasAMLCs and AMLC-branch NasAMLCs, respectively.

(F) Representative confocal micrographs showing μPASEs at *t* = 48 h stained for TFAP2C, NANOG, and SOX17, with IWP2 supplemented into the basal medium from *t* = 0 h.

In (C) and (F), experiments were repeated four times with similar results. Nuclei were counterstained with DAPI. Scale bars, 50 μm.
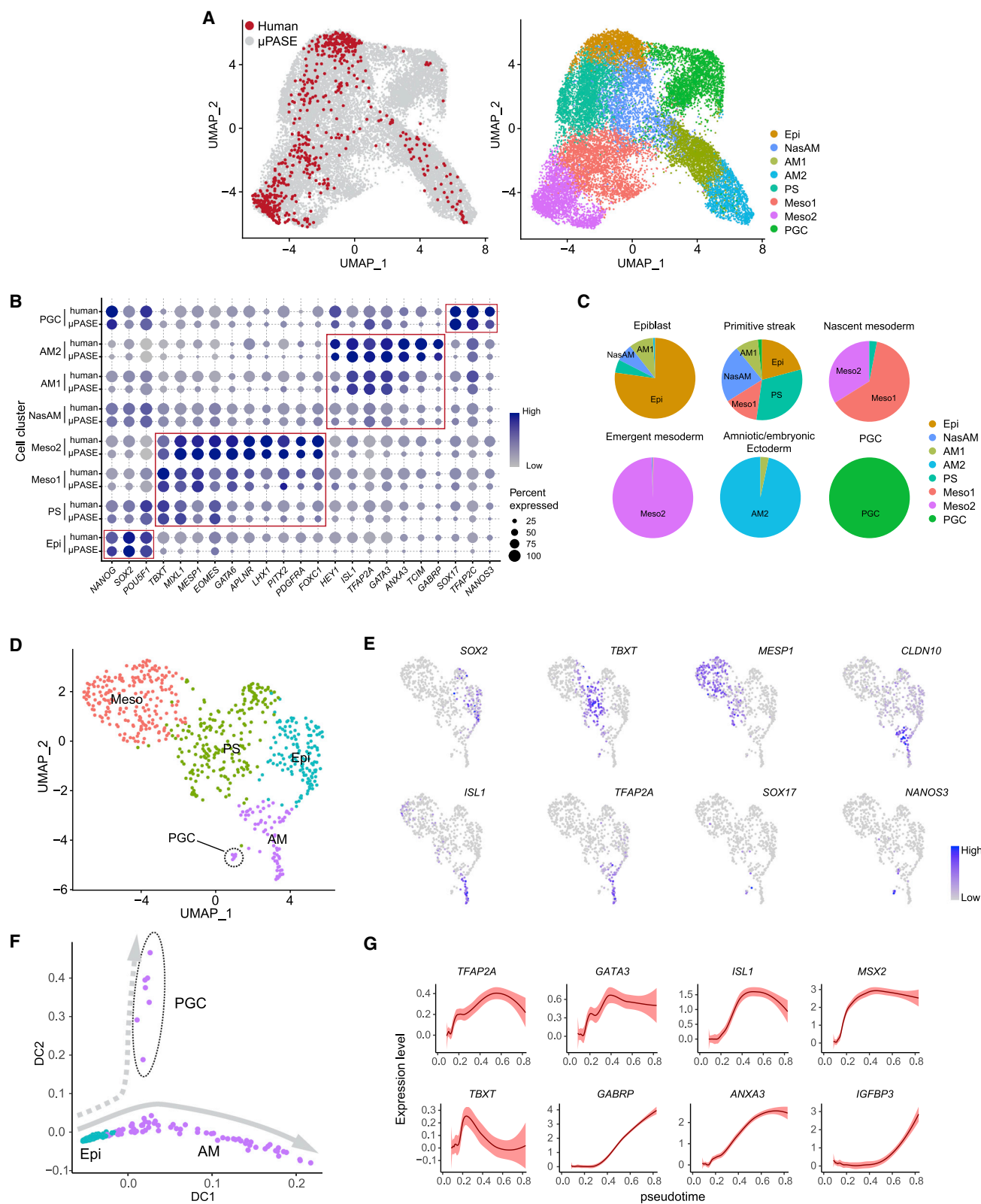
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and PGC seemingly display lineage bifurcation trajectories from the Epi (Figure 4F), similar to μPASEs in Figure 3A. In addition, as shown in Figure 4G, pseudotime gene expression dynamics of AM cells from the CS7 human gastrula are highly consistent with those of AMLCs in μPASEs in Figure 2D. However, because of the low number of PGCs, AM and PGC lineage bifurcation is not as evident as that shown in the μPASE diffusion map (Figure 3A), and we could not obtain pseudotime gene expression dynamics for PGCs.

To characterize pluripotency state transition in EpiLCs during the progressive development of the μPASE, we further performed scRNA-seq for cultured hPSCs and μPASEs at $t = 0$ h and $t = 12$ h. No distinguishable cell clusters or populations were observed when the scRNA-seq data were analyzed using Seurat R package (Mendeley Data Figure 1). Notably, although μPASEs at $t = 12$ h show upregulated *TFAP2A* expression owing to exogeneous BMP4 treatment, transcriptomes of cells in the μPASE have not yet changed sufficiently for the cells to emerge as distinct clusters in the UMAP. We then downsampled these datasets and compared them with transcriptome data of the human morula, pre-implantation epiblast from human blastocysts (Petropoulos et al., 2016), early post-implantation epiblast from days post-fertilization (d.p.f.) 9 and d.p.f. 11 *in vitro* cultured human embryos (Molè et al., 2021), and late post-implantation epiblast from the CS7 human gastrula (Tyser et al., 2021). As suggested by the PCA plot (Figures S4C and S4D), cultured hPSCs and EpiLCs in the μPASE share comparable transcriptome states and are both at a developmental stage between d.p.f. 11 and d.p.f. 16 epiblast. Thus, there was no compelling evidence showing heterogeneous pluripotency states or transitions at early stages of μPASE development.

Transcriptomic coordination of early development across different primate species

NHP monkey embryos, including cynomolgus (*Macaca fascicularis*) embryos, have been used successfully as an *in vivo* model to study primate development (Ma et al., 2019; Nakamura et al., 2016; Sasaki et al., 2016; Yang et al., 2021). Compared with NHP monkeys, chimpanzees share much more of human DNA (99%), making them our closest living relatives (Gibbs et al., 2007; Mikkelsen et al., 2005). Thus, we sought to generate μPASEs from chimpanzee induced pluripotent stem cells (iPSCs) to examine whether chimpanzee μPASEs (or C_μPASEs) would develop in a fashion similar to μPASEs made from hPSCs (or H_μPASEs). To this end, the same microfluidic protocol for H_μPASE development was used for C_μPASE formation. Under exogeneous hu-

man BMP4 stimulation, C_μPASEs also display prominent molecular and morphogenetic asymmetry, with the pole exposed to BMP4 differentiating progressively into squamous, flattened AMLCs positive for AM markers ISL1, GATA3, and TFAP2A and the opposite pole developing into germ layer lineages positive for TBXT and MIXL1 (Figure 5A). Similar to H_μPASEs, PGCLCs also emerge in C_μPASEs, as evidenced by positive immunostaining for SOX17, TFAP2C, NANOG, and BLIMP1 (Figure 5A). However, distinct from H_μPASEs, FOXA2⁺BLIMP1⁺ endoderm-like cells (EndoLCs) also emerge in C_μPASEs, and these EndoLCs randomly intermix with MeLCs and PGCLCs (Figure 5A). scRNA-seq analysis of C_μPASEs obtained at $t = 48$ h further confirms the presence of EndoLCs in C_μPASEs (Figures 5B and 5C). It is worth noting that C_μPASEs do not develop a prominent central lumen (the pro-amniotic-like cavity) as in H_μPASEs (Figure 5A).

Transcriptomic comparisons between related corresponding cell clusters in H_μPASEs, C_μPASEs, and human and cynomolgus embryos (Ma et al., 2019; Nakamura et al., 2016; Sasaki et al., 2016; Tyser et al., 2021; Yang et al., 2021) reveal that such cell clusters from these *in vitro* and *in vivo* systems, regardless of the Meso (Figure 5D), AM (Figure 5E), or PGC (Figure 5F) lineages, show strong correlations based on ontogenic genes identified from cynomolgus embryo transcriptome data (Nakamura et al., 2016; Sasaki et al., 2016; Yang et al., 2021). Consistently, when IWP2 was supplemented into C_μPASEs, developments of PSLCs and PGCLCs were inhibited, whereas ISL1⁺GATA3⁺ AMLCs still emerged, similar to IWP2-treated H_μPASEs (Figure 5G).

NODAL is essential for mesoderm development

Our DEG and pathway analyses of μPASE scRNA-seq data in Figures 2G, 2H, 3D–3F, S2F, and S2G suggest complex cell-cell interactions involved in early post-implantation human development. Thus, μPASE scRNA-seq data at $t = 24$ h were further analyzed using CellChat (Jin et al., 2021) for inference and analysis of ligand-receptor interactions (Figures S5A–S5C). Among cell clusters present in μPASEs at $t = 24$ h, AMLCs and PSLCs were identified as major sources of signaling ligands involved in key developmental pathways, such as BMP and WNT pathways (Ben-Haim et al., 2006; Bernardo et al., 2011; Clevers, 2006; Rivera-Perez and Magnuson, 2005; Wang et al., 2014; Zhao, 2003) (Figures S5A–S5C). This is consistent with a recent study of pre-gastrulation cynomolgus embryos (Yang et al., 2021) that shows that both AM and PS/Meso cells upregulate *BMP4* and *WNT5B* and extraembryonic mesenchyme cells

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Figure 4. Transcriptomic comparison between μPASEs and Carnegie stage 7 human gastrula

(A) Left: UMAP of integrated dataset of μPASEs from t = 24, 36, and 48 h (18,335 cells; gray) and CS7 human gastrula (647 cells, excluding irrelevant cells; red). Right: UMAP project of integrated dataset with cell identity annotations.

(B) Dot plot comparing expression of key marker genes across different cell clusters from μPASEs and CS7 human gastrula as indicated. The sizes and colors of dots indicate the proportion of cells expressing the corresponding genes and their averaged scaled values of log-transformed expression, respectively.

(C) Comparisons between human gastrula cell annotations in the original publication and annotations in the integrated dataset. The original annotations are indicated above the pie charts. See Mendeley Data Table 5.

(D) Re-analysis of related cells from CS7 human gastrula (647 cells). Cell identity annotations are color coded as indicated. Note that primordial germ cells (PGCs) appear in the AM cluster.

(E) Feature plots showing expression of selected lineage markers used for cell cluster annotations in (D).

(F) Trajectory inference (diffusion map) of the AM cluster (including PGCs) in (D).

(G) Expression dynamics (pseudotime) of selected genes during AM lineage development. Level of confidence (0.95) is indicated by band width.

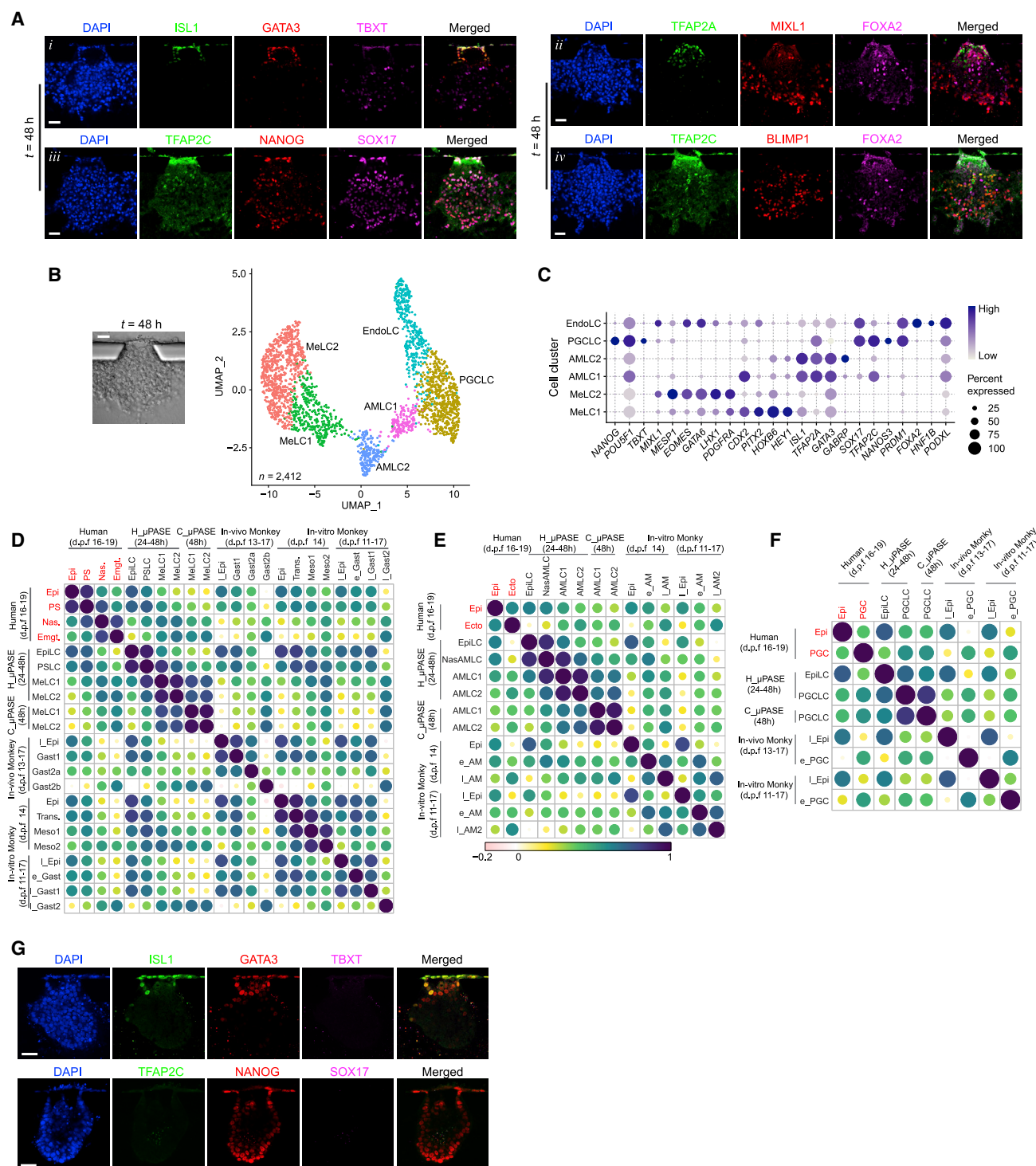Epi, Epiblast; PS, primitive steak; Meso, Mesoderm; AM, amniotic ectoderm; PGC, primordial germ cell.

Figure 5. Transcriptomic coordination of primate early post-implantation development

(A) Representative confocal micrographs showing chimpanzee μPASEs at t = 48 h stained for ISL1, GATA3, and TBXT (i); TFAP2A, MIXL1, and FOXA2 (ii); TFAP2C, NANOG, and SOX17 (iii); TFAP2C, BLIMP1, and FOXA2 (iv).

(B) Left: bright-field image showing a chimpanzee μPASE at t = 48 h. Right: UMAP plot of single-cell transcriptome dataset from chimpanzee μPASEs at t = 48 h. Cell identity annotations are color coded as indicated. n indicates the total cell number. EndoLC, Endoderm-like cell.

(C) Dot plot showing expression of key marker genes across the cell clusters as indicated. The sizes and colors of dots indicate the proportion of cells expressing the corresponding genes and their averaged scaled values of log-transformed expression, respectively.

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show high expression of *BMP2* and *BMP4* (Figures S5D–S5F). μPASEs lack extraembryonic mesenchyme cells, the effects of which might have been substituted for by exogenous supplementation of BMP4 in the μPASE protocol. For comparison, ligand-receptor interaction analysis was also conducted using CellChat on scRNA-seq data from E6.5 mouse embryos (Pijuan-Sala et al., 2019), revealing the extraembryonic ectoderm as the only source of BMP and WNT signals (Figures S5G–S5I). Interestingly, during the development of both μPASEs and pregastrulation cynomolgus embryos, non-canonical WNT pathways show a greater signaling strength than canonical WNT, whereas the opposite is observed for E6.5 mouse embryos (Figures S5A, S5D, and S5G).

Our ligand-receptor interaction analysis for μPASEs using CellChat further reveals that, compared to BMP and WNT pathways identified as an "incoming signal" for multiple cell clusters, NODAL signals appear to be perceived only by PSLCs (Figures S5A and S5B), suggesting a critical role of NODAL signaling in PSLC development. To study the functional role of NODAL, *NODAL*-knockout (KO) hPSC lines were generated and used for μPASE development (Figures S6A–S6E). By *t* = 48 h, the majority of cells in *NODAL*-KO μPASEs appear to have developed into flattened ISL1$^+$GATA3$^+$ AMLCs, and there are no cells disseminating away from the μPASE structure, in distinct contrast with wild-type control μPASEs (Figures 6A–6D and S6F). Consistently, scRNA-seq data obtained from *NODAL*-KO μPASEs at *t* = 48 h showed a dominant population of AMLCs at the expense of MeLCs (Figures 6B, 6C, and 6E–6G), further supporting the critical role of NODAL in MeLC development. PGCLCs were still evident in *NODAL*-KO μPASEs at *t* = 48 h, albeit with a much smaller percentage compared with those in wild-type controls (8.1% versus 28.2%; Figures 6D–6G and S6G).

We further conducted DEG and Gene Ontology (GO) enrichment analyses for PSLCs from *NODAL*-KO and wild-type control μPASEs (Figures 6H and 6I; Mendeley Data Table 7). In NODAL-KO PSLCs, BMP target genes, such as *ID2*, *TFAP2A*, and *ISL1*, are upregulated, whereas PS/Meso-related genes, such as *MESP1* and *MIXL1*, and EMT related genes, including *SNAI1* and *VIM*, are downregulated (Figure 6H). PI3K-Akt, WNT, and focal adhesion signaling pathways appear to be downstream targets of NODAL signaling in PSLCs (Figure 6I). We speculate

that lineage bifurcation between AMLCs and PSLCs from EpiLCs might be regulated by a competition between BMP and NODAL signaling. Absence or repression of NODAL signaling in the μPASE could lead to "hyper" BMP activities, which in turn cause excessive AMLC development (marked by an expanded ISL1 domain) and greater BMP activities in PSLCs. We also compared PGCLCs from wild-type and *NODAL*-KO μPASEs (Figure S6H and Mendeley Data Table 7). Consistently, when SB431542, a pharmacological NODAL signaling inhibitor, was supplemented into the microfluidic device, development of PSLCs/MeLCs in μPASEs was completely inhibited (Figure S6I). Impaired development of MeLCs in *NODAL*-KO μPASEs can be efficiently reversed by supplementing ACTIVIN A, a NODAL pathway agonist, into the channel opposite to BMP4 stimulation. In addition to rescuing MeLC development, supplementing ACTIVIN A to *NODAL*-KO μPASEs also leads to the specification of BLIMP1$^+$FOXA2$^+$ EndoLCs (Figure S6J). Supplementing wild-type μPASEs with ACTIVIN A also results in specification of BLIMP1$^+$FOXA2$^+$ EndoLCs (Figure S6K). We also successfully generated μPASEs using a different hPSC line, including its corresponding *NODAL*-KO line, and a different chimpanzee iPSC line and repeated IWP2 and SB431542 treatment assays with consistent results (Mendeley Data Figure 2). Consistently, we did not observe EndoLC development in μPASEs generated with these additional hPSC lines (Mendeley Data Figure 2).

Stringent criteria for distinguishing human trophoblast and amniotic ectoderm

In vivo, blastocyst trophoblast and pre-gastrulation AM both appear as flattened, squamous epithelium, and these two cell types share many lineage markers (Blakeley et al., 2015; Petropoulos et al., 2016; Tyser et al., 2021; Yang et al., 2021). Unique markers that can distinguish between these two cell types remain elusive, leading to confusion about true cell lineage identities in different human embryoids (Xu et al., 2002; Zhao et al., 2021). Through comparative transcriptome analysis of AM cells in the CS7 human gastrula (Tyser et al., 2021) and trophoblast cells in human blastocysts (Blakeley et al., 2015; Petropoulos et al., 2016), we identified a Trophoblast_Amnion ontogenic gene list (Figures S7A and S7B; Mendeley Data Table 8) and applied this list to examine the properties of human

(D) Heatmap of correlation matrix for primitive streak/mesoderm-related lineages, including those reported by others. Correlation coefficients between indicated cell types are calculated based on mesoderm ontogenic genes identified from cynomolgus embryo transcriptome data (131 in common). See Mendeley Data Table 6.

(E) Heatmap of correlation matrix for amnion-related lineages including those reported by others. Correlation coefficients between indicated cell types are calculated based on amnion ontogenic genes identified from cynomolgus embryo transcriptome data (142 in common). See Mendeley Data Table 6.

(F) Heatmap of correlation matrix for PGC-related lineages, including those reported by others. Correlation coefficients between indicated cell types are calculated based on PGC ontogenic genes identified for cynomolgus embryo transcriptome data (194 in common). See Mendeley Data Table 6.

(G) Representative confocal micrographs showing chimpanzee μPASEs at t = 48 h with IWP2 supplemented into the basal medium from t = 0 h, stained for ISL1, GATA3, and TBXT (top) and TFAP2C, NANOG, and SOX17 (bottom).

In (D)–(F), correlation coefficients are calculated using average ontogenic gene expression of single cells. Original cell annotations in published datasets are used. Log2 (reads per million +1) and Log2 (transcripts per million +1) are used for transcriptome datasets generated using 10× Chrome and Smart-seq2, respectively. Human, human gastrula from Tyser et al., 2021; Epi, epiblast; PS, primitive streak; Nas., Nascent mesoderm; Emgt., emergent mesoderm; Ecto., amniotic/embryonic ectoderm; PGC, primordial germ cell; H_μPASE and C_μPASE, μPASEs generated from human and chimpanzee cells, respectively; *in vivo* monkey, *in vivo* cynomolgus embryo from Nakamura et al., 2016 and Sasaki et al., 2016. L_EPI, late epiblast; Gast, gastrulation; e_PGC, early primordial germ cell; *in vitro* monkey, *in vitro* cultured cynomolgus embryo from Ma et al., 2019 and Yang et al., 2021. Trans, transition; Meso, mesoderm; l_EPI, late epiblast; e_Gast, early gastrulation; l_Gast, late gastrulation; e_AM, early amniotic ectoderm; l_AM, late amniotic ectoderm.

Expression of ontogenic genes is included in Mendeley Data Table 6. In (A) and (G), experiments were repeated four times with similar results. Nuclei were counterstained with DAPI. Scale bars, 50 μm.
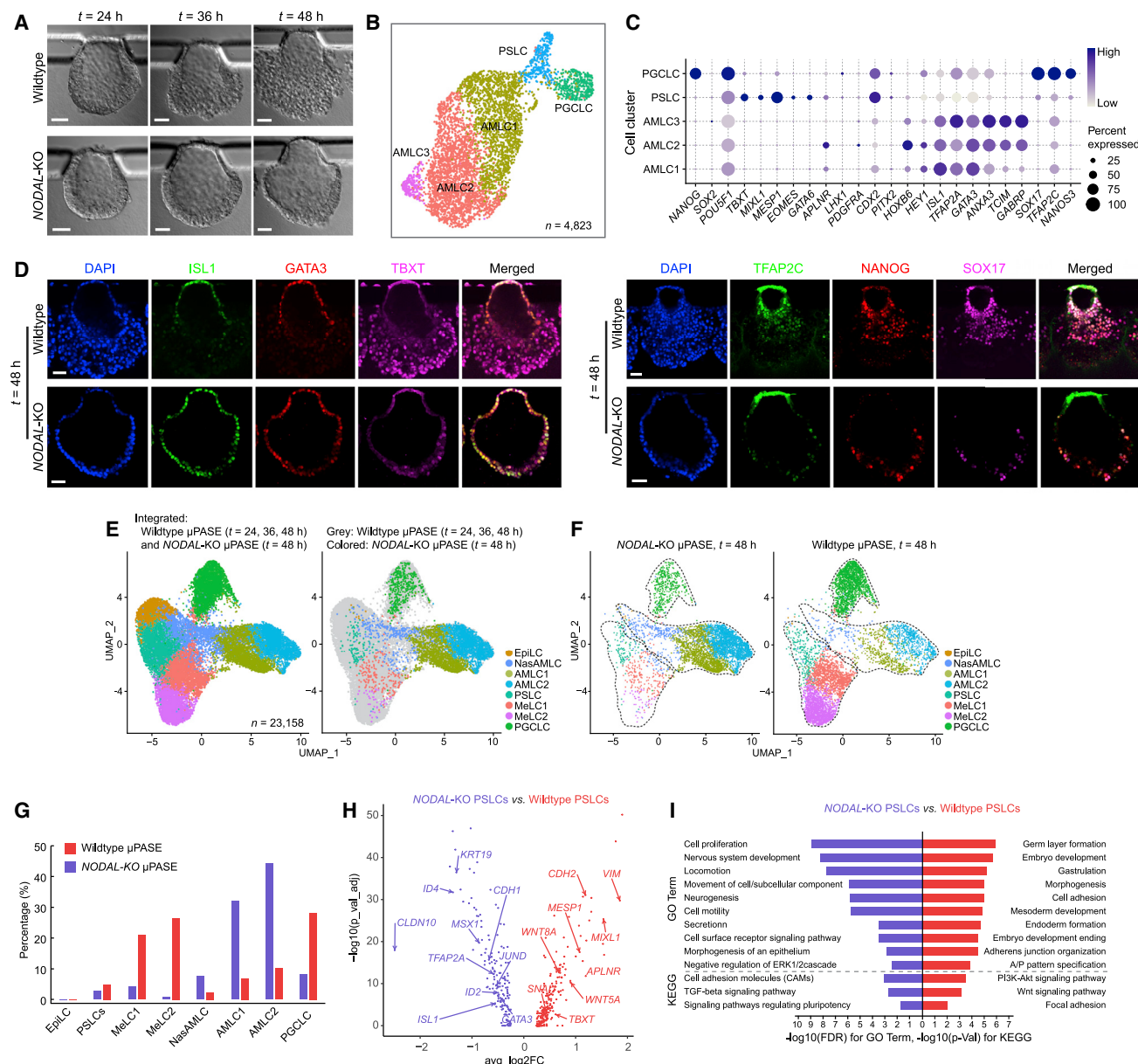
Figure 6. NODAL is essential for mesoderm development in μPASEs

(A) Representative bright-field images showing progressive development of μPASEs generated from wild-type (top) and NODAL-KO (bottom) hPSCs.

(B) UMAP plot of single-cell transcriptome data from NODAL-KO μPASEs at t = 48 h, with cell identity annotations color coded. n indicates the total cell number.

(C) Dot plot showing expression of key marker genes across the cell clusters in NODAL-KO μPASEs as indicated. The sizes and colors of dots indicate the proportion of cells expressing the corresponding genes and their averaged scaled values of log-transformed expression, respectively.

(D) Representative confocal micrographs showing μPASEs generated from wild-type and NODAL-KO hPSCs at t = 48 h, stained for ISL1, GATA3, and TBXT (left) and TFAP2C, NANOG, and SOX17 (right).

(E) Left: integrated UMAP plot of wild-type μPASEs at t = 24, 36, 48 h and NODAL-KO μPASEs at t = 48 h, color-coded according to cell identity annotations. Right: data from wild-type and NODAL-KO μPASEs are shown in different colors as indicated.

(F) UMAP plots of NODAL-KO (left) and wild-type (right) μPASEs at t = 48 h, with data isolated from (E). Dotted lines contour cell clusters corresponding to AMLC, PSLC/MeLC, and PGCLC lineages.

(G) Percentages of indicated cell types in wild-type (red) and NODAL-KO (blue) μPASEs.

(H) Volcano plots showing DEGs between PSLCs from wild-type and NODAL-KO μPASEs, with selected genes labeled.

(I) Enriched GO categories and KEGG pathways among DEGs between wild-type and NODAL-KO PSLCs.

In (D), experiments were repeated three times with similar results. Nuclei were counterstained with DAPI. Scale bars, 50 μm.
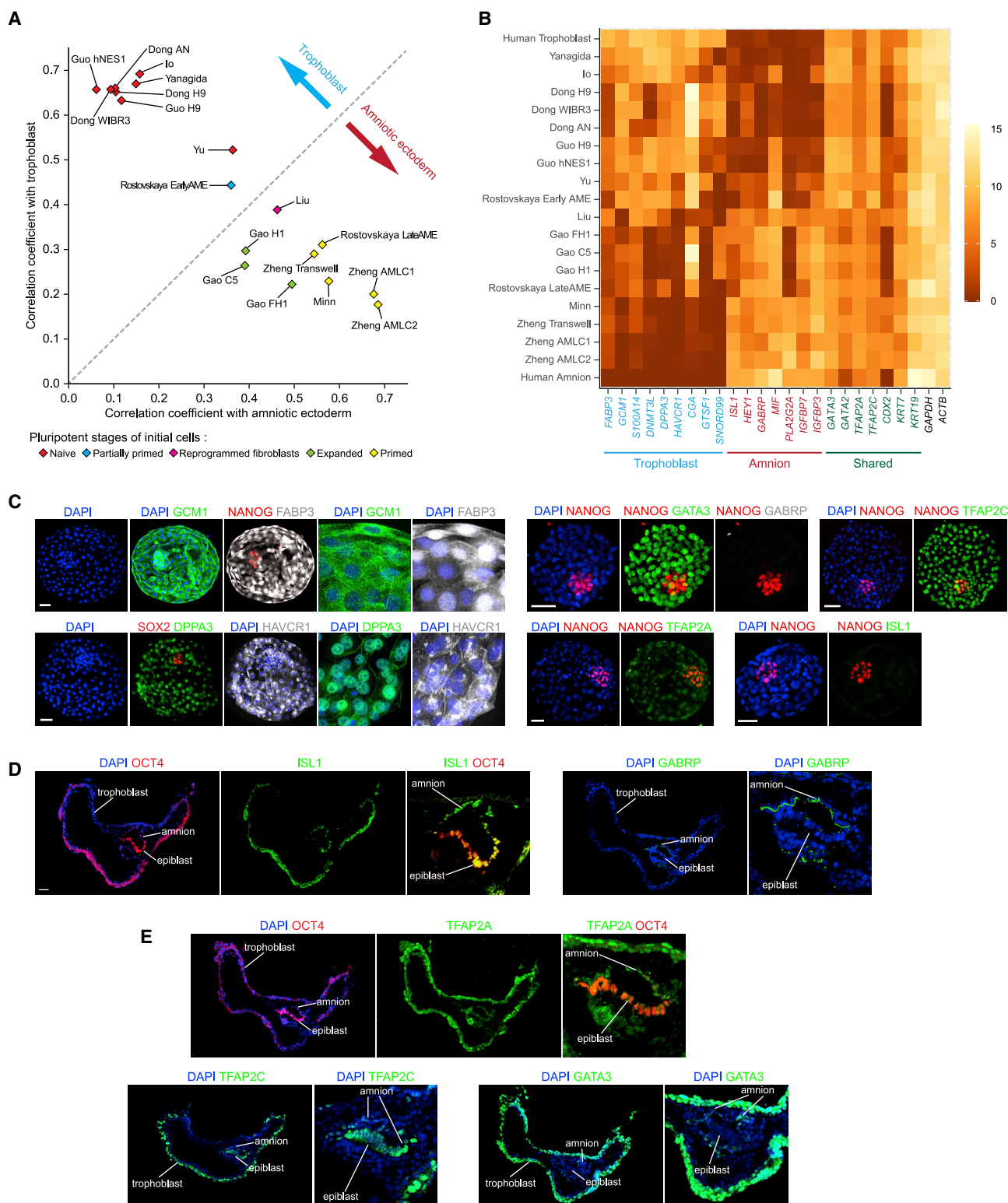
Figure 7. Stringent criteria for identifying human trophoblast and amniotic ectoderm

(A) Correlation coefficients of *in vitro*-derived cells with human trophoblast and amniotic ectoderm, computed using Trophoblast_Amnion ontogenic genes. See Mendeley Data Table 8.

(B) Heatmap showing expression levels of selected genes related to human trophoblast and amniotic ectoderm. Identified human trophoblast and amniotic ectoderm markers are highlighted in blue and red, respectively. Markers shared between human trophoblast and amniotic ectoderm are highlighted in green.

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trophoblast-like cells or AMLCs reported previously (Figure 7A). This ontogenic gene list contains several previously reported AM makers, such as *ISL1, GABRP,* and *IGFBP3* (Yang et al., 2021), and trophoblast makers, such as *GCM1, HAVCR1,* and *CGA* (Li et al., 2019; Okae et al., 2018; Pillai et al., 2019). Notably, several pan-pre-implantation embryo markers, including *DPPA3* and *DNMT3L* (Guo et al., 2021; Io et al., 2021; Yanagida et al., 2021), are found to be quite efficient markers for distinguishing between blastocyst trophoblast and pre-gastrulation AM. Based on quantified correlation coefficients between previously reported human trophoblast-like cells or AMLCs and human trophoblast or AM cells, we conclude that the transcriptome of BMP4-treated primed hPSCs, as in μPASEs, in 2D Transwell membrane-based AMLC differentiation assays (Zheng et al., 2019b) and in 2D-patterned gastrulation models (Minn et al., 2020), is similar to that of human AM cells, whereas the transcriptome of trophoblast-like cells derived from naive hPSCs (Dong et al., 2020; Guo et al., 2021; Io et al., 2021; Yanagida et al., 2021) is consistent with that of human blastocyst trophoblast. However, the transcriptome of trophoblast-like cells derived from extended pluripotent stem cells as well as that of trophoblast-like cells present in recently reported human blastocyst-like structures fall somewhat in between (Liu et al., 2021; Yu et al., 2021).

Using stringent criteria, we further identified a subset of DEGs within the Trophoblast_Amnion ontogenic gene list useful for distinguishing between human blastocyst trophoblast and pre-gastrulation AM and thus between human trophoblast-like cells and AMLCs (see STAR Methods; Figures 7B, S7A, and S7B). We should note that although AM transcriptome is obtained from the CS7 human gastrula (Tyser et al., 2021), which remains as the only *bona fide* human pre-gastrulation AM data currently available, most of the pre-gastrulation AM markers identified here, including *ISL1, HEY1, GABRP,* and *IGFBP3*, are upregulated in the pre-gastrulation AM of cynomolgus embryos (Yang et al., 2021).

We further conducted immunofluorescence analyses of IVF human blastocysts (D6), *in vitro* cultured cynomolgus embryos (D14), trophoblast stem cells (TSCs) (Okae et al., 2018), and AMLCs derived from BMP4-treated primed hPSCs (Zheng et al., 2019b) to ascertain these newly identified human blastocyst trophoblast and pre-gastrulation AM markers (Figures 7C–7E and S7C). Consistently, trophoblast cells of D6 human blastocysts show positive immunostaining for GCM1, FABP3, DPPA3, and HAVCR1 but are negative for ISL1 and GABRP (Figure 7C). In contrast, AM cells in D14 cynomolgus embryos show clear positive immunostaining for ISL1 and GABRP (Figure 7D).

GATA3, TFAP2A, and TFAP2C show positive immunostaining in both trophoblast cells of D6 human blastocysts and AM cells of D14 cynomolgus embryos (Figures 7C and 7E). Immunofluorescence analyses of TSCs (Okae et al., 2018), AMLCs (Zheng et al., 2019b), and μPASEs at *t* = 36 h further support the idea that ISL1, GCM1, and HAVCR1 can be utilized for distinguishing between trophoblast-like cells and AMLCs, whereas GATA3, TFAP2A, and TFAP2C could not (Figures S7C and S7D). We should note that ISL1 is a particularly useful nuclear marker for distinguishing between human blastocyst trophoblast and pre-gastrulation AM and thus between human trophoblast-like cells and AMLCs (Figures 7C–7E and S7C). The function of ISL1 in the AM in inducing Meso development in cynomolgus embryos has recently been documented (Yang et al., 2021).

DISCUSSION

Experimentation on human embryoids can lead to a better understanding of the mechanisms of human development and offers opportunities for functional genomic studies of disease-causing mechanisms, identification of therapeutic targets, and preclinical modeling of advanced therapeutics for precision medicine (Fu et al., 2021; Rossant and Tam, 2021). When developing new embryoids or improving existing embryoids, rigorous scientific scrutiny must be implemented to avoid incorrect interpretations or overstatements (Posfai et al., 2021; Zhao et al., 2021). Currently, there are few lineage markers accepted for distinguishing between human trophectoderm and AM cells, and most existing embryoids are still limited in imitating sequential cell lineage diversifications and 3D tissue organizations exhibited stereotypically in natural embryos. Besides using cell lineage markers, validation and authentication of embryoids are currently commonly conducted through comparative transcriptome studies (e.g., through integration of scRNA-seq datasets from embryoids and *in vivo* models) (Posfai et al., 2021; Zhao et al., 2021). However, caution should be taken when interpreting such integrated datasets because, although seemingly "unbiased," existing scRNA-seq data analysis tools still have notable artifacts depending on the cell types present, cell numbers and percentages of each cell type, and specific computational algorithms used in the analysis tools. For instance, AM cells only account for a very small cell population in post-implantation primate embryos; such a small AM cell number can cause significant issues when scRNA-seq data of AM cells are integrated with human blastocyst datasets wherein trophoblast cells are abundant. Thus, we propose that in order to validate cell identities in peri-implantation human embryoids,

(C) Representative confocal micrographs showing E6 human blastocysts, stained for GCM1, NANOG, and FABP3; SOX2, DPPA3, and HAVCR1 (images on the right show magnified views of the trophectoderm); NANOG, GATA3, and GABRP; NANOG and TFAP2C; NANOG and TFAP2A; and NANOG and ISL1.
(D) Representative confocal micrographs showing *in vitro* cultured D14 cynomolgus embryos, stained for OCT4 and ISL1 and GABRP. Images on the right show magnified views of the epiblast and amniotic ectoderm.
(E) Representative confocal micrographs showing *in vitro* cultured D14 cynomolgus embryos, stained for OCT4 and TFAP2A, and TFAP2C and GATA3. Images on the right show magnified views of the epiblast and amniotic ectoderm.
In (A) and (B), transcriptome data are obtained from: human trophoblast (Blakeley et al., 2015; Petropoulos et al., 2016), Human amnion (Tyser et al., 2021), Zheng Transwell (Zheng et al., 2019b), Gao C5, Gao H1, Gao FH1 (Gao et al., 2019), Minn (Minn et al., 2020), Liu (Liu et al., 2021), Yu (Yu et al., 2021), Guo H9, Guo hNES1 (Guo et al., 2021), Io (Io et al., 2021), Yanagida (Yanagida et al., 2021), Dong H9, Dong WIBR3, Dong AN (Dong et al., 2020), Rostovskaya EarlyAME, and Rostovskaya LateAME (Rostovskaya et al., 2022).
Correlation coefficients and gene expression heatmap were calculated based on averages of experimental repeats, if any. When the published work contains transcriptome datasets from multiple time points, datasets from the experimental endpoint are utilized.
In (C)–(E), experiments were repeated twice with similar results. Nuclei were counterstained with DAPI. Scale bars, 50 μm.

expression of key cell fate markers needs to be clearly demonstrated, in addition to transcriptome comparison based on scRNA-seq data. The cell identity markers established in this work for distinguishing between human trophoblast and AM cells represent one step in this important direction.

LIMITATIONS OF STUDY

As demonstrated in this study, μPASEs provide a promising tractable experimental model for exploring previously inaccessible phases of early post-implantation human development. However, μPASEs lack a few key embryonic and extraembryonic lineages that present in the post-implantation human embryo, including the hypoblast (or extraembryonic endoderm), extraembryonic mesoderm, and trophoblast cells. Therefore, μPASEs only recapitulate certain aspects of the lineage diversification and development of the posterior end of the embryonic sac during early post-implantation human development. Additionally, μPASEs disintegrate and lose their embryonic-like structure soon after 48 h in culture, as differentiating MeLCs that are undergoing EMT emigrate from the μPASE structure, which limits the potential of μPASEs for prolonged culture to investigate cell lineage development and embryonic tissue formation at later developmental stages. It should also be noted that cell type annotations and lineage inference analyses of the μPASE in this work are solely based on transcriptomic studies. Thus, caution should be taken when interpreting the data included in this resource.

STAR★METHODS

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SUPPLEMENTAL INFORMATION

Supplemental information can be found online at https://doi.org/10.1016/j.stem.2022.08.009.

AUTHOR CONTRIBUTIONS

Y.Z. and J.F. conceived and initiated the project. Y.Z. designed and performed microfluidic embryoid experiments and conducted scRNA-seq data analyses and interpretations. R.Z.Y. conducted CellChat analysis and participated in data analysis. S.S. participated in microfluidic embryoid experiments and repeated experiments with additional cell lines. M.K. and T.S. established and characterized *NODAL*-KO hPSC lines. L.X., Y. Li, and T.L. performed human embryo experiments. R.Y., A.G., W.J., Y.N., and K.R.C. performed monkey embryo experiments. X.X., S.N.E., Y. Liu, and A.M.R.I. helped with microfluidic embryoid experiments. W.W. helped with scRNA-seq data analyses. Y.Z. and J.F. wrote the manuscript. J.F. supervised the study. All authors edited and approved the manuscript.

DECLARATION OF INTERESTS

Two patents related to this work have been filed (US20190321415/WO2018106997 by J. Fu, Y. Zheng, and S.N. Esfahani; US2020049721/PCT/US20/49721 by J. Fu and Y. Zheng).

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STAR★METHODS

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
Mouse anti-ISL1 antibody	DSHB	Cat# 39.4D5, RRID:AB_2314683
Rabbit anti-GATA3 antibody	Cell Signaling Technology	Cat# 5852, RRID:AB_10835690
Mouse anti-GATA3 antibody	Thermo Fisher Scientific	Cat# MA1-028, RRID:AB_2536713
Rabbit anti-GABRP antibody	Thermo Fisher Scientific	Cat# PA5-46830, RRID:AB_2610501
Rabbit anti-MIXL1 antibody	Atlas Antibodies	Cat# HPA005662, RRID:AB_2667382
Rabbit anti-HEY1 antibody	Abcam	Cat# ab154077, RRID:AB_2893447
Mouse anti-TFAP2A antibody	Santa Cruz Biotechnology	Cat# sc-12726, RRID:AB_667767
Goat anti-TBXT antibody	Thermo Fisher Scientific	Cat# PA5-46984, RRID:AB_2610378
Rabbit anti-NANOG antibody	Cell Signaling Technology	Cat# 4903, RRID:AB_10559205
Goat anti-NANOG antibody	R and D Systems	Cat# AF1997, RRID:AB_355097
Mouse anti-TFAP2C antibody	Santa Cruz Biotechnology	Cat# sc-12762, RRID:AB_667770
Goat anti-SOX17 antibody	R&D Systems	Cat# AF1924, RRID:AB_355060
Rabbit anti-BLIMP1 antibody	Cell Signaling Technology	Cat# 9115, RRID:AB_2169699
Rabbit anti-FOXA2 antibody	Cell Signaling Technology	Cat# 8186, RRID:AB_10891055
Mouse anti-OCT4 antibody	Santa Cruz Biotechnology	Cat# sc-5279, RRID:AB_628051
Rabbit anti-OCT4 antibody	Santa Cruz Biotechnology	Cat# sc-9081, RRID:AB_2167703
Rabbit anti-SOX2 antibody	Millipore	Cat# AB5603, RRID:AB_2286686
Mouse anti-DPPA3 antibody	Millipore	Cat# MAB4388, RRID:AB_2094156
Mouse anti-GCM1 antibody	Santa Cruz Biotechnology	Cat# sc-101173, RRID:AB_2108121
Goat anti-HAVCR antibody	R and D Systems	Cat# AF1750, RRID:AB_2116561
Rabbit anti-FABP3 antibody	Proteintech	Cat# 10676-1-AP, RRID:AB_2102309
Goat anti-GATA6 antibody	R and D Systems	Cat# AF1700, RRID:AB_2108901
Mouse anti-NODAL antibody	Abcam	Cat# ab55676, RRID:AB_2151660
Mouse anti- GAPDH antibody	Sigma-Aldrich	Cat# G8795, RRID:AB_1078991
HRP-conjugated anti-mouse IgG antibody	Santa Cruz Biotechnology	Cat# sc-516102, RRID:AB_2687626
Donkey anti-Rabbit 546	Thermo Fisher Scientific	Cat# A10040, RRID:AB_2534016
Donkey anti-Mouse 488	Thermo Fisher Scientific	Cat# A-21202, RRID:AB_141607
Donkey anti-Goat 647	Thermo Fisher Scientific	Cat# A-21447, RRID:AB_2535864
Donkey anti-Rabbit 568	Thermo Fisher Scientific	Cat# A10042, RRID:AB_2534017
Donkey anti-Mouse 647	Thermo Fisher Scientific	Cat# A-31571, RRID:AB_162542
Goat anti-Mouse IgG1 488	Thermo Fisher Scientific	Cat# A-21121, RRID:AB_2535764
Goat anti-Mouse IgG2b 647	Thermo Fisher Scientific	Cat# A-21242, RRID:AB_2535811
Biological samples		
Human embryo	First People's Hospital of Yunnan Province	N/A
Cynomolgus monkeys (*Macaca fascicularis*)	Yunnan Key Laboratory of Primate Biomedical Research (LPBR) in China. Astrid Fagræus laboratory of the Karolinska Institutet in Sweden.	N/A
Chemicals, peptides, and recombinant proteins		
Epidermal growth factor (EGF)	R&D System	Cat# 236-EG-200
CHIR99021	Cayman Chemical	Cat# 13122
A83-01	Tocris Bioscience	Cat# 2939
SB431542	STEMCELL Technologies	Cat# 72234
Y27632	Tocris	Cat# 1254
FGF2	Thermo Fisher Scientific	Cat# PHG0261

Continued

REAGENT or RESOURCE	SOURCE	IDENTIFIER
BMP4	R&D Systems	Cat# 314-BP-050
IWP2	Tocris	Cat# 3533
LY294002	Selleckchem	Cat# S1105
ACTVIN A	R&D Systems	Cat# 338-AC-050
Critical commercial assays		
LookOut Mycoplasma PCR Detection Kit	Sigma-Aldrich	Cat# MP0035-1KT
Click-iT™ Plus EdU Cell Proliferation Kit	Thermo Fisher Scientific	Cat# C10637
Deposited data		
Data files for RNA sequencing	Gene Expression Omnibus	GSE185643
Processed dataset	Gene Expression Omnibus	GSE185643, https://umichibbl.shinyapps.io/shinyapp/
Raw data for plots	Mendeley Data	https://doi.org/10.17632/cvv5rzymhd.1
Experimental models: Cell lines		
Human: H9 (WA09) (female)	WiCell	NIH registration number: 0062
Human: ESI-017 (male)	BioTime, Inc.	NIH registration number: 0093
Chimpanzee: Chimpanzee iPSC lines (C3651, male and C4955, female)	Gallego Romero et al., 2015; Pavlovic et al., 2018	N/A
Human: Human trophoblast stem cells (hTSCs)	Okae et al., 2018	N/A
Human: ESI-017 NODAL-KO hPSC line	Chhabra et al., 2019	N/A
Oligonucleotides		
NODAL_crRNA_1: 5′-AGGCUCAGCAUG UACGCCAG-3′	Thermo Fisher Scientific	N/A
NODAL_crRNA_2: 5′-AGACAUCAUCCG CAGCCUAC-3′	Thermo Fisher Scientific	N/A
NODAL exon 1 Forward Primer: 5′-CTTC CTTCTGCACGCCTGGTGG-3′	Thermo Fisher Scientific	N/A
NODAL exon 1 Reverse Primer: 5′-CCAA CCCACAGCACTTCCCGAG-3′	Thermo Fisher Scientific	N/A
Recombinant DNA		
pCXLE-EGFP expression plasmid	Addgene	plasmid # 27082; RRID: Addgene_27082
Software and algorithms		
R	Open source	https://www.r-project.org
Python	Open source	https://www.python.org/
ImageJ	Open source	https://imagej.net/software/fiji/
Other		
mTeSR™1	STEMCELL Technologies	Cat# 85,850
mTeSR™ Plus	STEMCELL Technologies	Cat# 100-0276
Gibco™ Geltrex™ LDEV-Free, hESC-Qualified, Reduced Growth Factor Basement Membrane Matrix	Thermo Fisher Scientific	Cat# A1413302
Corning™ Matrigel™ hESC-Qualified Matrix	Thermo Fisher Scientific	Cat# 08-774-552
KnockOut™ Serum Replacement	Thermo Fisher Scientific	Cat# 10828028
Insulin-Transferrin-Selenium-Ethanolamine (ITS -X)	Thermo Fisher Scientific	Cat# 51500056
Essential 6 medium (E6)	Thermo Fisher Scientific	Cat# A1516401

RESOURCE AVAILABILITY

Lead contact
Further information and requests for resources and reagents should be directed to and will be fulfilled by the lead contact, Jianping Fu (jpfu@umich.edu).

Materials availability
Requests for NODAL-knockout H9 hPSCs should be addressed to Yi Zheng and Jianping Fu (e-mail: yzheng88@syr.edu, jpfu@umich.edu).

Data and code availability
scRNA-seq data have been deposited at the Gene Expression Omnibus and are publicly available as of the date of publication. Accession numbers are listed in the key resources table. Interactive visualization of processed datasets is available at https://umichibbl.shinyapps.io/shinyapp/.

　This paper does not report original codes.

　Any additional information required to reanalyze the data reported in this paper is available from the lead contact upon request. Raw data for the plots were deposited on Mendeley at https://doi.org/10.17632/cvv5rzymhd.1.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Cell culture
Cell lines used in this study are listed in the key resources table. Culture/growth conditions are described in Method details.

Human embryos
IVF human blastocysts (D6) were obtained from couples who had at least one healthy baby after *in vitro* fertilization with signed consents. Human embryo experiments were approved by the Medicine Ethics Committee of the First People's Hospital of Yunnan Province (KHLL2020-KY064).

Monkey embryos
IVF cynomolgus macaque embryos were obtained from healthy female cynomolgus monkeys aged 5 to 8 years old. All animals were housed either at the facility of the Yunnan Key Laboratory of Primate Biomedical Research (LPBR) in China, or at the Astrid Fagræus laboratory of the Karolinska Institutet in Sweden. Both facilities are accredited by AAALAC international. Experimental protocols for using cynomolgus macaque embryos were approved by the Institutional Animal Care and Use Committee of LPBR in China (KBI K001115033/01,01) and by the Jordbruksverket in Sweden (Ethical Permit Number N277/14).

METHOD DETAILS

Ethics statement
μPASE lacks both the primitive endoderm and trophoblast and thus cannot form the yolk sac or placenta. Therefore, μPASE does not have human organismal form or potential. Furthermore, μPASE disassembles and lose structural integrity after $t = 48$ h, and all experiments were terminated by no later than $t = 72$ h. All protocols with hPSCs were approved by the Human Pluripotent Stem Cell Research Oversight Committee at the University of Michigan, Ann Arbor. Protocols of human embryo experiments were approved by the Medicine Ethics Committee of the First People's Hospital of Yunnan Province (KHLL2020-KY064). This Medicine Ethics Committee has 13 members, including lawyers, scientists and clinicians with relevant expertise. This committee evaluated the scientific merits and ethical justification of human embryo experiments and conducted a full review of the donation and use of human embryo samples. All human embryos donated to this study were surplus frozen embryos from couples who had at least one healthy baby after *in vitro* fertilization. All donor couples signed informed consents for voluntary donations at the Department of Reproductive Medicine in the First People's Hospital of Yunnan Province. No economic benefits were offered during the process. Couples were informed that their embryos would be used for experimental studies of human development and that their donation would not affect their *in vitro* fertilization processes.

Cell culture
H9 (WA09, WiCell; NIH registration number: 0062, female) and ESI-017 (BioTime, Inc.; NIH registration number: 0093, male) hPSCs and chimpanzee iPSC lines (C3651, male and C4955, female) (Gallego Romero et al., 2015; Pavlovic et al., 2018) were maintained in a feeder-free culture system using mTeSR medium (STEMCELL Technologies). For hPSCs, culture plate was coated with 1% lactate dehydrogenase-elevating virus (LDEV)-free, hESC cell-qualified reduced growth factor basement membrane matrix Geltrex (Thermo Fisher Scientific; derived from Engelbreth-Holm-Swarm mouse tumors) before cell seeding. For chimpanzee iPSCs, culture plate was coated with 1% Matrigel (Thermo Fisher Scientific; extracted from Engelbreth-Holm-Swarm mouse sarcoma). Cells were visually examined during each passage to ensure absence of spontaneously differentiated, mesenchymal-like cells in culture. Cells were

used before P70. Both human and chimpanzee cells have been authenticated by original sources as well as in-house by immunostaining for pluripotency markers and successful differentiation into the three germ layers. Cells were maintained for at least ten passages and authenticated as karyotypically normal. Karyotype analysis was performed by Cell Line Genetics. Both human and chimpanzee cell lines were tested negative for mycoplasma contamination (LookOut Mycoplasma PCR Detection Kit, Sigma-Aldrich).

Generation of *NODAL*-knockout hPSCs

To generate *NODAL*-knockout (KO) hPSCs, a 58-bp portion of genomic DNA within NODAL exon 1 was deleted by CRISPR/Cas9 using two crRNA purchased from Thermo Fisher Scientific [NODAL_crRNA_1: 5′-AGGCUCAGCAUGUACGCCAG-3′; NODAL_crRNA_2: 5′-AGACAUCAUCCGCAGCCUAC-3′] (Figures S7A–S7C). Deplexes of crRNA:tracrRNA were prepared using a standard protocol and introduced into H9 hPSCs with the Cas9 enzyme and the pCXLE-EGFP expression plasmid (a gift from Shinya Yamanaka; Addgene plasmid # 27,082; RRID: Addgene_27082) for constitutional expression of EGFP using the NEON electroporation system (Thermo Fisher Scientific). EGFP-expressing single cells were collected and seeded onto Matrigel-coated 96-well plates by fluorescence-activated cell sorting (FACSAria Fusion, BD Biosciences) with CloneR single-cell culture supplement diluted with mTeSR Plus medium (STEMCELL Technologies). To detect the anticipated deletion, genomic DNA was isolated from single-cell derived clones and subjected to PCR using the following primers designed for amplification of NODAL exon 1 [Forward Primer: 5′-CTTCCTTCTGCACGCCTGGTGG-3′; Reverse Primer: 5′-CCAACCCACAGCACTTCCCGAG-3′]. Resulting amplicons were subjected to Sanger sequencing using a primer 5′-CTTCCTTCTGCACGCCTGGTGG-3′. ESI-017 *NODAL*-KO hPSC line is a generous gift from Aryeh Warmflash at Rice University (Chhabra et al., 2019).

Western blotting

Wildtype and *NODAL*-KO hPSCs were exposed to GSK3 inhibitor CHIR99021 (CHIR, 10 μM; Cayman Chemical) for 24 h to augment expression of NODAL protein before cells were lysed with RIPA buffer containing the cOmplete protease inhibitor cocktail (Roche). Protein concentration was determined by the Bradford assay using Protein Assay Dye Reagent Concentrate (Bio-Rad). An equal amount of protein (80 μg) from cell lysates of widetype and *NODAL*-KO hPSCs was resolved on 10% SDS-PAGE and transferred onto PVDF membranes (Thermo Fisher Scientific). Immunostaining was performed by blocking PVDF membranes with 5% skim milk for 1 h at room temperature followed by incubation overnight at 4°C with mouse monoclonal antibodies to human NODAL (Abcam ab55676; 1:500 dilution) or human GAPDH (Sigma-Aldrich G8795; 1:20,000 dilution) diluted in 5% skim milk. Membranes were washed with PBS (PBS) containing 0.1% Tween 20 (PBS-T) and stained with HRP-conjugated anti-mouse IgG antibody (Santa Cruz sc-516102; 1:3000 dilution) diluted in 5% skim milk for 1 h at room temperature. Protein bands were detected by a chemiluminescence assay using Super-Signal West Pico Plus and Super-Signal West Femto reagents (Thermo Fisher Scientific).

Cynomolgus macaque

Healthy cynomolgus monkeys (*M. fascicularis*), aged from 5 to 8 years old, were used in this study. All animals were housed either at the facility of the Yunnan Key Laboratory of Primate Biomedical Research (LPBR) in China, or at the Astrid Fagræus laboratory of the Karolinska Institutet in Sweden. Both facilities are accredited by AAALAC international. Experimental protocols for using cynomolgus macaque embryos were approved by the Institutional Animal Care and Use Committee of LPBR in China (KBI K001115033/01,01) and by the Jordbruksverket in Sweden (Ethical Permit Number N277/14). Animals involved in this study were never used for other treatments.

In vitro fertilization and culture of cynomolgus macaque embryo

In vitro fertilized cynomolgus macaque embryos were generated as described previously (Niu et al., 2014). Briefly, healthy female cynomolgus monkeys aged 5 to 8 years old with regular menstrual cycles were selected as oocyte donors. Cynomolgus monkeys were treated with recombinant human follicle stimulation hormone (Merck, Gonal-f) for 8 days, followed by administration of recombinant human chorionic gonadotropin (Merck, Ovidrel) on day 9. After 32–35 h, oocytes were collected by laparoscopic follicular aspiration. Metaphase II (MII) oocytes were used for intracytoplasmic sperm injection to generate zygotes, and fertilization was confirmed by the presence of two pronuclei. Zygotes were cultured in embryo culture medium-9 (polyvinyl alcohol (0.1 mg/mL), calcium chloride (1.9 mM), magnesium chloride (0.46 mM), potassium chloride (3.0 mM), sodium chloride (113.8 mM), sodium bicarbonate (25.0 mM), sodium lactate (4.5 mM), Minimum Essential Medium (MEM) amino acid, MEM nonessential amino acid, and gentamicin (10 mg/mL)) containing 10% fetal calf serum in 37°C incubator supplied with 5% CO_2 until the blastocyst stage. *In vitro* culture of NHP monkey blastocysts beyond the implantation stage has been described previously (Yang et al., 2021). In brief, frozen NHP monkey blastocysts were thawed using Thawing Media (Kizatato) and cultured in blastocyst culture medium (Origio) for at least 4 h to recover. Blastocysts were then treated with Acidic Tyrode's solution (Sigma) to remove zona pellucida before being transferred onto an ibiTreat 8-well μ-plate (Ibidi) containing 300 μL of pre-equilibrated *in vitro* culture medium 1 (advanced DMEM/F12, 20% FBS, l-Glutamine, l-cysteine, 1x Penicillin/Streptomycin, 1x ITS-X, supplemented with beta-estradiol, progesterone). On the second day, 150 μL of culture medium was aspirated, before 200 μL of pre-equilibrated *in vitro* culture medium 2 (advanced DMEM/F12, 30% KSR, l-Glutamine, l-cysteine, 1x Penicillin/Streptomycin, 1x ITS-X, supplemented with beta-estradiol, progesterone) was added into the ibiTreat 8-well μ-plate. Embryo growth was recorded daily, and culture medium was replenished every two days until Day 14.

NHP monkey embryo cryosection and immunocytochemistry

Day 14 NHP monkey embryos were fixed using 2% paraformaldehyde (PFA; buffered in PBS) overnight at 4°C before being washed with PBS. Fixed embryos were dehydrated by 30% sucrose overnight at 4°C before embedded in Tissue-Tek O.C.T. compound (Sakura) and frozen in liquid nitrogen. Frozen blocks were used for cryosection using CryoStar NX70 Cryostat (Thermo Fisher Scientific) according to manufacturer's protocol. Immunofluorescence staining was then performed according to a standard procedure. Briefly, sections were thawed and air-dried at room temperature. After washing with PBS, sections were incubated in blocking buffer (3% FBS) diluted in PBS with 0.1% Triton X-100) for 1 h at room temperature and then incubated overnight at 4°C with primary antibodies diluted in blocking buffer. The sections were washed with PBS-T and incubated with secondary antibodies diluted in blocking buffer for 2 h at room temperature. After washing thoroughly with PBS-T, sections were mounted and imaged. Secondary antibodies were used in dilution of 1:500. Images were acquired by a Zeiss 700 LSM Confocal Microscope and analyzed by iMaris.

Immunocytochemistry for D6 human embryos

D6 human embryos were fixed with 4% PFA for 20 min at room temperature in a 96-well plate, washed with PBS, and then permeabilized and blocked by 0.2% Triton X-100 supplemented with 3% BSA overnight at 4°C. Embryos were then transferred onto a new well with primary antibodies for 16–18 h at 4°C. Embryos were washed 3 times with PBS containing 0.05% Tween 20, 15 min every time, before being transferred to secondary antibody solutions for 4 h at room temperature. Embryos were then washed 3 times in PBS containing 0.05% Tween 20, 15 min every time, before being transferred onto an ibiTreat 8-well μ-plate. All antibodies were diluted by 1% BSA solutions.

Generation of μPASEs

The microfluidic device was fabricated by bonding a polydimethylsiloxane (PDMS) structure layer to a glass coverslip. Singly dissociated hPSCs or chimpanzee iPSCs were loaded into the cell loading channel and guided to settle into preformed, concave Geltrex pockets by tilting the device by 90° for 10 min hPSCs were then maintained in mTeSR containing 10 μM Y27632 (Tocris) for 18 h to allow for cell clustering. At $t = 0$ h, culture medium in all medium reservoirs of the device was switched to a fresh basal medium comprising Essential 6 medium (E6; Thermo Fisher Scientific) and FGF2 (20 ng mL^{-1}; Thermo Fisher Scientific), with additional 50 ng mL^{-1} BMP4 (R&D Systems) supplemented only in the cell loading channel. A detailed protocol for microfluidic device fabrication and generation of μPASEs can be found elsewhere (Zheng et al., 2019b, 2021). To examine possible involvements of different signaling pathways involved in the development of μPASEs, IWP2 (10 μM; Tocris), SB431542 (10 μM; STEMCELL Technologies) or LY294002 (20 μM; Selleckchem) was supplemented into basal medium from $t = 0$ h. For assays with ACTIVIN A, 100 ng mL^{-1} ACTIVIN A (R&D Systems) was supplemented into the channel opposite to BMP4 stimulation from $t = 0$ h. For *in situ* proliferation measurements, Click-iT EdU Kit (Invitrogen) was used as per manufacturer's protocol. Diluted EdU solution (10 μM) was introduced into all reservoirs of the device at $t = 24$ h. After 3 h of incubation, the μPASEs was fixed and imaged.

Immunocytochemistry for μPASEs

μPASEs were fixed with 4% PFA for 12 h before being permeabilized in 0.1% SDS solution (SDS dissolved in PBS) for 3 h μPASEs were then blocked in 4% donkey serum (Sigma-Aldrich) at 4 °C for 24 h, followed by incubation with primary antibody solutions at 4 °C for another 24 h. Samples were then labeled with donkey-raised secondary antibodies (1:500 dilution) at 4 °C for 24 h. 4′,6-diamidino-2-phenylindole (DAPI; Thermo Fisher Scientific) was used for counterstaining cell nuclei. Both primary and secondary antibodies were prepared in 4% donkey serum supplemented with 0.1% NaN$_3$. 70 μL antibody solutions were added to each medium reservoir of the microfluidic device for immunostaining. All primary and secondary antibodies used in this study are listed in key resources table. All confocal micrographs of μPASEs were acquired by a NIKON A1SI Confocal Microscope equipped with a photomultiplier tube (PMT) detector and processed using ImageJ 1.53c.

Quantification of SOX17⁺ cells

μPASEs at $t = 48$ h with or without LY294002 were stained for SOX17 using the protocol described above. All confocal micrographs captured at the central focal plane of structure (50 μm above the microfluidic device bottom surface) were used for quantification. The percentage of SOX17 + cells was calculated as the ratio between the area of SOX17 channel and DAPI channel for each μPASE. The binarization and measurement of the images was conducted using ImageJ 1.53c. Quantitative results were analyzed using independent, two-tailed Student's t test in Excel (Microsoft). $p < 0.05$ was considered statistically significant.

Trophoblast stem cells and derivation of AMLCs using BMP4

Human trophoblast stem cells (hTSCs) derived from blastocysts were generously provided by Dr. H. Okae and Dr. T. Arima (Okae et al., 2018). hTSCs were maintained in 1% Geltrex coated 6-well plates in DMEM/F12 supplemented with 0.05 mM 2-mercaptoethanol, 0.2% fetal bovine serum (FBS), 0.5% knockout serum replacement (KSR), 0.5% penicillin-streptomycin, 0.3% BSA, 1% ITS-X supplement, 1.5 μg mL^{-1} L-ascorbic acid, 50 ng mL^{-1} epidermal growth factor (EGF), 2 μM CHIR99021, 1 μM A83-01, 1 μM SB431542, 0.8 mM valproic acid (VPA) and 5 μM Y27632, and passaged using TrypLE.

   To obtain AMLCs by treating hPSCs with BMP4, singly dissociated hPSCs were suspended in mTesR1 containing 10 μM Y27632 and seeded in a 1% Geltrex coated 6-well plate at a density of 2.5×10^3 cells cm^{-2}. Note that this cell seeding density is optimized to avoid extensive cell death (when cell density is too low) or emergence of PSLCs in culture through an unspecified inductive effect of

incipient AMLCs (Zheng et al., 2019b). 18 h after cell seeding, culture medium was switched to Essential 6 medium containing FGF2 (20 ng mL^{-1}) and BMP4 (50 ng mL^{-1}). Resulting AMLCs were fixed and stained after 48 h of BMP4 treatment.

Single cell dissociation and RNA-sequencing

μPASEs (H9) at different time points were washed twice with DMEM/F12 for 10 min and incubated with Accutase for 1 h. After incubation, μPASEs in the microfluidic device were dissociated into single cells by gentle agitating. Single cells from six microfluidic devices were collected and pooled into PBS containing 0.5% BSA before being centrifuged at 300 g for 5 min. The resultant cell pellet was re-suspended in PBS containing 0.5% BSA. Within 1 h after cell dissociation, cells were loaded into the 10X Genomics Chromium system (10X Genomics). 10X Genomics v.3 libraries were prepared according to the manufacturer's instructions. Libraries were then sequenced using paired-end sequencing with a minimum coverage of 20,000 raw reads per cell using an Illumina NovaSeq-6000. scRNA-seq data were aligned and quantified using Cell Ranger Single-Cell Software Suite (v.3.1.0, 10X Genomics) against the *Homo sapiens* (human) genome assembly GRCh38.p13 from ENSEMBL. Chimpanzee μPASEs (C3651) were dissociated and sequenced following the same protocol. scRNA-seq data from chimpanzee μPASEs were aligned against Pan_tro_3.0 from ENSEMBL.

Data integration, dimensionality reduction, and clustering

Analysis of scRNA-seq data and integration of scRNA-seq datasets were performed using Seurat R package (v.4.0.0.0, https://satijalab.org/seurat/) (Butler et al., 2018; Satija et al., 2015). Default setups were used unless noted otherwise. Briefly, a single batch of scRNA-seq dataset was filtered based on total number of genes detected and percentage of total mitochondrial genes. Gene expression was then calculated by normalizing the raw count with the total count before multiplying by 10,000 and log transformed. Cell cycle was regressed out based on cell cycle scores (CellCycleScoring) during data scaling process (ScaleData). PCA analysis (RunPCA) was then performed on filtered data followed by embedding into low dimensional space with Uniform Manifold Approximation and Projection (UMAP; RunUMAP). Identification of cell clusters by a shared nearest neighbor (SNN) modularity optimization-based clustering algorithm was achieved using the FindClusters function in the Seurat R package. For integration of different scRNA-seq datasets, count matrices of different datasets were filtered and normalized separately before being integrated using the IntegrateData function based on 2,000 anchor features. After integration, the integrated scRNA-seq dataset was analyzed following the standard Seurat pipeline. Annotation of cell clusters was based on expression of canonical lineage marker genes.

RNA velocity analysis and partition-based graph abstraction analysis (PAGA)

Bam files generated by the Cell Ranger pipeline was used for RNA velocity analysis. Genome annotations GRCh38.p13 were used for counting spliced and unspliced mRNA of individual cells. Python package scVelo (v.0.2.2, https://scvelo.readthedocs.io) was employed to perform RNA velocity analysis using dynamical modeling (scv.tl.velocity) (Bergen et al., 2020; La Manno et al., 2018). Function 'scv.pl.velocity_embedding_stream' was used to project RNA velocities onto UMAP plots or diffusion maps. All default parameters were used unless noted otherwise. Python package Scanpy (v.1.8.0, https://scanpy.readthedocs.io/en/stable/) was used for evaluating the relationship between different cell clusters by the partition-based graph abstraction (PAGA) analysis (Wolf et al., 2018). Briefly, Seurat object was converted to "Loom" file and passed to Scanpy. Neighborhood graph of observations was then computed using 20 PCs (sc.pp.neighbors). Finally, PAGA graph was plotted with a "eq_tree" layout.

Gene regulatory network analysis

Regulatory activity of transcription factors associated with specific cell types was assessed using the R-package SCENIC (Single Cell rEgulatory Network Inference and Clustering, v.1.1.2–2, https://github.com/aertslab/SCENIC) (Aibar et al., 2017). Briefly, regulatory modules were first identified by inferring co-expression with transcription factors using GENIE3. Each co-expression module was then analyzed using *cis*-regulatory motif analyses (RcisTarget). Only modules with significant motif enrichment of the correct upstream regulator were retained. The human motif collection v9 and the cisTarget databases for hg38 were used in the pipeline (https://resources.aertslab.org/cistarget/). Filtered counts of the integrated Seurat object were used as input of SCENIC. All default parameters were used in SCENIC unless noted otherwise.

Trajectory interference using diffusion map and pseudotime

Diffusion maps were obtained by the R-package Density (https://bioconductor.org/packages/release/bioc/html/destiny.html), which computes kernel density estimates with parametric starts and asymmetric kernels (Angerer et al., 2016). To generate three-dimensional diffusion maps, PCA embeddings of the integrated Seurat object were used as input of 'diffmap' function. Cells in the three-dimensional diffusion map were color-coded consistently with the UMAP plot of the integrated Seurat object. Three-dimensional visualization was realized using R-package rgl (3D Visualization Using OpenGL). For trajectory inference and pseudotime analysis of specific cell lineages, relevant cell clusters of the integrated Seurat object were extracted using Subset function (Seurat). PCA embeddings of selected cell clusters were then used as the input of 'diffmap' function. To visualize gene expression dynamics, diffusion pseudotime (dpt) was utilized. The roots of the diffusion map were automatically chosen by the algorithm; and the EpiLC cluster was arbitrarily made as the initial cell type. Expression levels of selected genes were fitted as a function of the pseudotime with "loess" method by using 'geom_smooth' function in ggplot2 R package (v.3.3.3). Cells with extreme dpt values were counted as outliner and excluded from gene expression dynamic analysis. RNA velocity vectors were computed as described above and superimposed onto

the diffusion maps. In the case of lineage bifurcation, branching point and branches of different lineages were identified by locally fitting half-lines to single-cell data in the diffusion map using K-Branches R package (https://github.com/theislab/kbranches) (Chlis et al., 2017).

Differential gene expression, gene ontology enrichment, and pathway analyses

Gene expression data depicted in feature plots and dot plots in this paper were calculated from raw counts after NormalizeData function followed by ScaleData function of the Seurat package, unless noted otherwise. Differentially expressed genes (DEGs) between different cell types were identified using FindMarkers function (Seurat), with a minimal fold difference of 0.25 in the logarithmic scale and >10% detection rate in either of the two cell types under comparison. Gene ontology (GO) enrichment and pathway analyses were performed using online tool iPathwayGuide (Advaita Bioinformatics) referencing AmiGO Gene Ontology database and KEGG PATHWAY Database, respectively.

Integration, re-analysis, and PCA of published human embryo data

The scRNA-seq dataset of the CS7 human gastrula was integrated with μPASE scRNA-seq datasets. Since μPASEs contain AMLCs, PGCLCs and PSLCs/MeLCs, only relevant cell types from the human gastrula (Amniotic/embryonic ectoderm, Epiblast, Primitive Streak, Nascent Mesoderm, Emergent Mesoderm) were extracted according to the annotations in the original publication. Human gastrula scRNA-seq dataset was generated using Smart-seq2. To compare with μPASE scRNA-seq datasets generated using the 10X Genomics Chromium system, raw counts of each cell in the human gastrula were normalized to exon sizes before being utilized to create the Seurat object (*input count = raw count/exon size* × 1,000). Exon size information was obtained from GRCh38.p13, ENSEMBL. To integrate scRNA-seq data from the human gastrula with those of μPASEs, μPASE scRNA-seq datasets from the three time points (t = 24, 36 and 48 h) were first filtered. Normalization, Scaling (including cell cycle regression) and PCA were then performed separately on each of the datasets after which they were combined using the reciprocal PCA approach (IntegrateData function, Seurat) based on 30 dimensions and 2,000 anchor features. Dimensionality reduction and clustering were then performed as described above. Downsampling of large scRNA-seq datasets was performed using Seurat R package (Subset), which randomly selects 100 cells from every cluster in the original datasets to form a new Seurat object. This step helps prevent larger datasets from dominating the downstream analysis.

For re-analysis of human gastrula scRNA-seq dataset, relevant cell types were processed with the default Seurat pipeline. Trajectory inference using diffusion map and pseudotime analysis was achieved using the same pipeline as for μPASEs, as descried earlier. For PCA, large datasets were first downsampled; then the normalized counts of epiblast ontogenetic genes (Mendeley Data Table 5) from different datasets were merged into a matrix. The matrix was scaled to ensure each gene has zero average and unit variance. PCAs was calculated and visualized using Seurat R package.

Cross-species comparison

Cross-species comparisons utilize scRNA-seq datasets from the CS7 human gastrula, human μPASEs (t = 24, 36, and 48 h), chimpanzee μPASEs (t = 48 h), *in vivo* cynomolgus embryos (D13 – 17), and *in vitro* cultured cynomolgus embryos (D11 – 17) (Ma et al., 2019; Nakamura et al., 2016; Sasaki et al., 2016; Tyser et al., 2021; Yang et al., 2021). To mitigate artifacts resulted from different sequencing platforms, for datasets generated using Smart-seq2, raw counts were first normalized to exon length, as described above. Exon size information of cynomolgus monkey was obtained from Macaca_fascicularis_6.0, ENSEMBLE. Average raw counts normalized by exon length for specific cell types in different datasets were then transformed into log2(counts per million reads +1) and used as gene expression levels for calculating the correlation coefficient matrix and heatmap plotting.

Ontogenic genes of mesoderm and PGC lineages were identified through DEG analysis of *in vivo* cynomolgus embryo datasets. Briefly, for mesoderm lineage ontogenic genes, DEGs of "Gast1", "Gast2a" and "Gast2b" as compared to "PostL-EPI" were identified using FindMarkers function (Seurat) (Nakamura et al., 2016). All DEGs, with a minimal fold difference of 2 in the logarithmic scale (logfc), >10% detection rate in either of the two cell types under comparison, and adjusted p-value < 0.05, were used as mesoderm lineage ontogenic genes (Mendeley Data Table 6). To identify PGC lineage ontogenic genes, DEGs of "ePGC" as compared to "PostL-EPI"(Sasaki et al., 2016), with a minimal fold difference of 2 in the logarithmic scale (logfc), >10% detection rate in either of the two cell types under comparison, and adjusted p-value < 0.05, were used as PGC lineage ontogenic genes (Mendeley Data Table 6). Amniotic ectoderm ontogenic genes were obtained through DEG analysis of Day 14 *in vitro* cultured cynomolgus embryos (Yang et al., 2021). Specifically, DEGs of "early-Amnion" and "late-Amnion" as compared to "EPI" were identified using FindMarkers function (Seurat). All DEGs, with a minimal fold difference of 0.75 in the logarithmic scale (logfc), >10% detection rate in either of the two cell types under comparison, and adjusted p-value < 0.01, were used as amniotic lineage ontogenic genes (Mendeley Data Table 6).

Analysis of cell-cell interactions

R package CellChat was used to perform cell-cell communication analysis (http://www.cellchat.org/) (Jin et al., 2021). Briefly, based on manually curated databases that consider known structural compositions of ligand-receptor interactions, CellChat infers and analyzes intercellular communication networks from scRNA-seq data using network analysis and pattern recognition. Seurat object including count matrix and clustering results from each dataset is imported to CellChat. The default human database was used for μPASE and cynomolgus embryo dataset analyses, whereas the default mouse database was used for mouse embryo dataset

analysis. For both databases, only secreted signaling pathways from Kyoto Encyclopedia of Genes and Genomes (KEGG) were used. Default values were used for all parameters, except that the truncated mean was lowered to 5% to increase algorithm sensitivity.

Comparisons between human trophoblast and amniotic ectoderm

Several published RNA-seq datasets related to human blastocyst trophoblast and amniotic ectoderm were surveyed in this work. Datasets generated from bulk RNA-sequencing or Smart-seq2 were first normalized by exon sizes as described earlier. Specifically, blastocyst trophoblast cells annotated in (Blakeley et al., 2015; Petropoulos et al., 2016) were utilized as the reference of human blastocyst trophoblast. Both two datasets were analyzed in this work to mitigate batch effects and systematic variations. Amniotic/embryonic ectoderm cells from the CS7 human gastrula were utilized as the reference of human amniotic ectoderm (Tyser et al., 2021). DEGs between trophoblast and amniotic ectoderm were first identified using trophoblast cells from (Blakeley et al., 2015; Petropoulos et al., 2016), respectively, as compared to amniotic/embryonic ectoderm cells from the CS7 human gastrula, with a minimal fold difference of 0.75 in the logarithmic scale (logfc) and adjusted p-value < 0.05. Overlapped DEGs identified from both comparative transcriptome analyses were used as Trophoblast_Amnion ontogenic genes (n = 299) to calculate correlation coefficients of published datasets as well as AMLCs from this work with human blastocyst trophoblast and amniotic/embryonic ectoderm (Mendeley Data Table 8). Correlation coefficients and gene expression heatmap were calculated based on averages of experimental repeats, if any. The average gene expression level of the two trophoblast datasets was used to calculate correlation coefficients of surveyed cells with the human blastocyst trophoblast. To identify the most reliable markers for distinguishing trophoblast cells from amniotic ectoderm cells, more stringent criteria were applied than those for identifying Trophoblast_Amnion ontogenic genes. Among Trophoblast_Amnion ontogenic genes with logfc >2, only the genes that are negative in the other cell type (normalized count <0.05) were identified as reliable markers. Overlapped markers revealed for both trophoblast datasets were proposed as markers for distinguishing human trophoblast cells from amniotic ectoderm cells.

QUANTIFICATION AND STATISTICAL ANALYSIS

Quantitative results were analyzed using independent, two-tailed Student's t test in Excel (Microsoft). p < 0.05 was considered statistically significant.