Tissue microenvironment dictates the state of human induced pluripotent stem cell-derived endothelial cells of distinct developmental origin in 3D cardiac microtissues

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ABSTRACT

Each tissue and organ in the body has its own type of vasculature. Here we demonstrate that organotypic vasculature for the heart can be recreated in a three-dimensional cardiac microtissue (MT) model composed of human induced pluripotent stem cell (hiPSC)-derived cardiomyocytes (CMs), cardiac fibroblasts (CFs) and endothelial cells (ECs). ECs in cardiac MTs upregulated expression of markers enriched in human intramyocardial ECs (iECs), such as *CD36, CLDN5, APLNR, NOTCH4, IGFBP3, ARHGAP18,* which were previously identified in the single-cell RNA-seq dataset from the human fetal heart (6.5-7 weeks post coitum). We further show that the local microenvironment largely dictates the organ-specific identity of hiPSC-derived ECs: we compared ECs of different developmental origins derived from two distinct mesoderm subtypes (cardiac and paraxial mesoderm) and found that independent of whether the ECs were cardiac or paraxial mesoderm derived, they acquired similar identities upon integration into cardiac microtissues. This was

confirmed by single-cell RNA-seq. Overall, the results indicated that whilst the initial

gene profile of ECs was dictated by developmental origin, this could be modified by

the local tissue environment such that the original identity was lost and the organotypic

identity acquired through local environmental signals. This developmental "plasticity"

in ECs has implications for multiple pathological and disease states.

INTRODUCTION

Development of the vascular system is one of the earliest events in organogenesis

and defects in this process often result in embryonic and postnatal lethality.

Endothelial cells (ECs) that form the inner lining of blood and lymphatic vessels are

specialized cells that adapt to local microenvironmental cues to support the function

of various organs. The heart is a prime example of the importance that interplay

between the vasculature and the myocardium has in organ growth, remodeling, and

function. ECs in the heart originate from several developmental lineages that later

converge to similar states depending on their location. Structurally and functionally,

ECs in the heart can be divided into the endocardium, intramyocardial capillary ECs,

coronary arteries/veins and lymphatic ECs. During development, heart ECs

predominantly originate in the lateral plate mesoderm that includes both pre-cardiac

and cardiac mesoderm (Milgrom-Hoffman et al., 2011). Endocardium and *sinus*

venosus are the two major sources on intramyocardial and coronary ECs (He and

Zhou, 2018; Sharma et al., 2017a, 2017b; Tian and Zhou, 2022). Furthermore, genetic

ablation of the *sinus venosus* EC lineage results in compensation from the endocardial

EC lineage (Sharma et al., 2017b). Endocardium- and *sinus venosus*-derived EC

progenitors, converge to an increasingly similar state that is dictated by the local

microenvironment, despite being initially transcriptionally distinct (Phansalkar et al.,

2021). In addition, recent studies showed that cardiac lymphatic ECs predominately

originate from two lineages, namely Isl1+ second heart field and Pax3+ paraxial

mesoderm (PM) (Lioux et al., 2020; Stone and Stainier, 2019), which further increases

the spectrum of developmental origins of ECs in the heart.

Endothelial cells in the intramyocardial capillaries, or iECs, constitute a specialized

barrier between the blood and the myocardium. They deliver oxygen and essential

nutrients, specifically fatty acids, to the cardiomyocytes to fulfill high energy demands
of the working myocardium (Brutsaert, 2003). Endocardial ECs (eECs) form the lining
of the inner surface in the ventricles and atria and play important roles during the
development of working myocardium, such as formation of trabeculae network and
compaction of the myocardium (Qu et al., 2022). Analysis of developing mouse heart
identified markers that distinguish eECs and iECs (Iso et al., 2018; Zhang et al.,
2016a). Recent single-cell RNA sequencing (scRNA-seq) studies of the human fetal
heart confirmed that some of the markers identified in mouse are conserved in human,
including *CD36* and *FABP5* for iECs and *NPR3* and *CDH11* for eECs, among others
(McCracken et al., 2022; Miao et al., 2020).

Human pluripotent stem cells (hPSCs) represent a valuable *in vitro* model to study
early stages of human development, including the heart (Hofbauer et al., 2021a). Over
the past several years, protocols to differentiate cardiac cell types from hPSCs, such
as different sub-types of cardiomyocytes, epicardial cells, cardiac fibroblasts and ECs
were developed (Thomas et al., 2022). Engineered multicellular cardiac tissues can
be generated by combining these different cell types (Campostrini et al., 2021). These
have proved useful to investigate the contribution of non-cardiac cell types to
cardiomyocyte cell maturation and disease (Giacomelli et al, 2020a). In addition,
methods to create multicellular cardioids from hPSCs have been developed to model
embryonic stages from self-organized cardiac and foregut structures to early
morphogenesis during heart tube formation (Drakhlis et al., 2021; Hofbauer et al.,
2021b; Lewis-Israeli et al., 2021; Silva et al., 2021). Recent studies showed that by
following a specialized developmental program, hPSCs can be differentiated into ECs
from several developmental lineages, such as extra- and intraembryonic hemogenic
ECs (Ditadi et al., 2015; Ng et al., 2016; Uenishi et al., 2014), endocardial ECs
(Mikryukov et al., 2021) and liver sinusoidal ECs (Gage et al., 2020). We previously
developed a method to co-differentiate cardiomyocytes and ECs from hPSCs from
cardiac mesoderm (Giacomelli et al., 2017). These hPSC-derived ECs expressed a
number of cardiac specific genes like *MEOX2*, *GATA4*, *GATA6* and *ISL1*, while tissue
specific endocardial or intramyocardial markers were still absent, likely because of the
lack of local microenvironmental cues.

Recently, we established a three-dimensional (3D) cardiac microtissue (MT) model
composed of ECs, cardiomyocytes and cardiac fibroblasts, all derived from hiPSCs
(Giacomelli et al., 2020). We showed that hPSC-derived cardiomyocytes in cardiac 3D
MTs showed enhanced functional and structural maturation via interaction with ECs
and cardiac fibroblasts. We further showed that developmental origin of the fibroblasts
was critical in this model, as cardiac-, but not skin-, fibroblasts supported
cardiomyocyte maturation in 3D MTs.
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However, whether 3D cardiac MTs actually induce organ-specific characteristics and
to a what extent the developmental origin of ECs plays a role has not been investigated.
We therefore aimed here to address these questions by comparing ECs derived from
two distinct mesoderm sub-types: MESP1+ cardiac mesoderm and PAX3+ paraxial
mesoderm. To do this, we utilized our earlier protocol to differentiate cardiac
mesoderm-derived ECs (Giacomelli et al., 2017) and developed a new protocol to
differentiate ECs from paraxial mesoderm. Cardiac MTs were then generated using
these two sources of ECs. Although newly differentiated ECs from the two origins
showed distinct identities, they strikingly became more similar after extended culture
in cardiac MTs. Furthermore, based on eEC- and iEC-specific signatures extracted
from a published scRNA-seq dataset of human fetal heart (Asp et al., 2019), we
observed an iEC rather than an eEC identity for both developmental origins after MT
culture. In summary, this study shows that although certain characteristics are
inherited from progenitors, ECs are "plastic" and efficiently adapt to the
microenvironment to acquire new tissue-specific signatures. Our results provide new
insights into how organ/tissue-specific cell identities are acquired; this will inform the
preparation of hiPSC -derived, organ specific ECs for disease modeling and drug
development but as importantly, will provide a platform for understanding how EC
plasticity might be regulated by microenvironmental context.
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128 **RESULTS**
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130 **Differentiation of endothelial cells from cardiac and paraxial mesoderm**
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We set out to derive ECs from human induced pluripotent stem cells (hiPSCs) via both

cardiac and paraxial mesoderm intermediates. To obtain ECs from cardiac mesoderm

(CMECs), we used a protocol established previously in our group (Giacomelli et al.,

2017) (Figure 1A). Briefly, BMP4 (20 ng/ml), Activin A (ACTA, 20 ng/ml) and

CHIR99021 (CHIR, 1.5 µM) were used to induce cardiac mesoderm from day 0 till day

3. XAV-939 (XAV, 5 µM) and VEGF (50 ng/ml) were used to induce CMECs and early

cardiomyocytes (CMs) from day 3 to day 6. For paraxial mesoderm-ECs (PMECs), we

adapted a protocol developed by Loh et al. (Loh et al., 2016)(Figure 1B). Briefly, high

CHIR (8 µM) was used for the first two days followed by XAV (5 µM) for one day to

induce posterior presomitic mesoderm (pPSM) on day 3 and low CHIR (1.5 µM) was

used to induce anterior presomitic mesoderm (aPSM) from day 3 to day 5. VEGF was

added to induce PMECs from day 5 to day 6. In order to characterize paraxial

mesoderm differentiation and the PMEC lineage we established a double fluorescent

hiPSC reporter line (NCRM1 PAX3^{Venus}MSGN1^{mCherry}) (Figure S1A). CRISPR/Cas9

assisted gene editing was used to target the fluorescent protein Venus to the PAX3

locus leading to transcriptional control by the endogenous PAX3 regulatory elements.

The MSGN1^{mCherry} reporter was generated using a BAC construct integrated into the

cells using the piggyBac transposon system. Flow cytometry analysis at different

stages of paraxial mesoderm differentiation showed efficient induction of pPSM

(MSGN-mCherry positive cells) and aPSM (PAX3-Venus positive cells) on day 2-3

and day 5, respectively (Figure S1B). Using our paraxial mesoderm protocol, more

than 70% of cells acquired MSGN1-mCherry expression on day 2-3 (Figure S1C) and

more than 50% of cells acquired PAX3-Venus on day 6 (Figure S1D). We next

confirmed induction of endogenous PAX3 protein expression by immunostaining with

a PAX3-specific antibody (Figure S1E). Notably, more than 90% of cells were positive

for PAX3 on day 5; this could be because of relatively weak endogenous expression

of Venus that could not be detected in PAX3^{low} cells.

Gene expression analysis further confirmed comparable expression of pan-mesoderm

markers *TBXT* and *MIXL1* in both cardiac and paraxial mesoderm differentiation

conditions. On the other hand, expression of cardiac genes (*MESP1*, *GATA4* and

NKX2-5) was restricted to cardiac mesoderm differentiation conditions and expression

of paraxial mesoderm genes (*MSGN1*, *TBX6*, *PAX3*) was restricted to paraxial
mesoderm differentiation conditions (Figure S1F).

Both cardiac and paraxial mesoderm differentiation conditions resulted in comparable
percentages of CD144+CD140a- ECs on day 6 and day 8 of differentiation (Figure
1C-D). CD144+CD140a- ECs were sorted on day 6 and day 8 of differentiation from
both cardiac and paraxial mesoderm conditions and underwent RNA sequencing
(RNA-seq). Principle component analysis (PCA) showed that CMECs and PMECs
clustered separately along PC1, and day 6 and day 8 were separated along PC2
(Figure 1E). On day 6, 3307 and 2592 genes were significantly differentially
upregulated (FDR<0.05, fold-change>2) in CMECs and PMECs respectively (Table
S1). Gene ontology (GO) analysis showed that cardiac related genes were specifically
upregulated in day 6 CMECs (CMEC_D6), while genes related to skeletal system
development and function were specifically upregulated in day 6 PMECs (PMEC_D6)
(Figure 1F, Table S2). Genes involved in heart development, like *GATA4*, *GATA5*,
TBX3, *ISL1* and *MYH6*, were highly expressed in day 6 and day 8 CMECs. *TBX3*,
ISL1 and *MYH6* were upregulated from day 6 to day 8 in CMECs (Figure 1G).
Essential genes for skeletal muscle development like *PAX3*, *TBX1*, *FOXC1*, *EYA1* and
MEOX1 were largely expressed in day 6 and day 8 PMECs. *FOXC1* and *EYA1* were
upregulated from day 6 to day 8, while *TBX1* and *MEOX1* were downregulated (Figure
1G). In summary, unbiased expression analysis by bulk RNA-seq showed differential
gene expression signatures of cardiac and paraxial mesoderm derived ECs that
corresponded to their known expression profiles *in vivo*.

Reconstruction of the differentiation trajectories of ECs by single-cell RNA-seq

Having demonstrated that the two differentiation protocols result in ECs with distinct
characteristics, we undertook an unbiased analysis of the complete cell population
and reconstructed the EC differentiation trajectories. To this end, we performed
scRNA-seq on day 6 of CMEC and PMEC differentiation from two independent
biological replicates (Figure 2A-B, Figure S2A-B). The replicates appeared highly
similar in a low-dimensional representation (Figure S2C) and were therefore combined
for further analysis. Any remaining, undifferentiated hiPSCs were excluded from

further analysis (Figure S2D-E). In the CMEC differentiation data set, cells were
grouped into 3 clusters (cardiac mesoderm, cardiomyocytes and CMECs), as
established previously (Cao et al., 2022) (Figure 2C). The three cell types were
identified by known marker genes (Figure 2D, S3A-B, Table S3). The cardiac
mesoderm cluster was characterized by mesoderm and early cardiac genes, such as
MESP1, SMARCD3, ABLIM1, TMEM88, ISL1, MYL5, as well as the cell cycle-related
genes *CDK6* and *NEK2*. The CMEC cluster was characterized by EC markers (*CDH5,*
CD34, KDR, HEY2, TEK, TIE1, ACVRL1, SOX17, ENG, ICAM2, PECAM1).
Cardiomyocytes were identified by expression of cardiomyocyte-associated genes,
including *MYL4, TNNI1, MYL7, ACTA2, TNNT2, HAND2* and *NKX2-5*. To reveal the
differentiation trajectories of the cells, we calculated the diffusion pseudotime using a
cardiac cell mesoderm cell as root (Figure 2E). Pseudotime increased towards
CMECs and cardiomyocytes, suggesting that both cell types differentiated from a
common cardiac mesoderm progenitor.
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In the PMEC differentiation data set, all cells were divided into 3 clusters (Figure 2F),
which were interpreted as being paraxial mesoderm, PMECs and mixed lateral plate
mesoderm (LPM)/sclerotome using marker gene analysis (Figure S3C, D, Table S4).
The paraxial mesoderm cluster was characterized by expression of aPSM and
dermomyotome genes, such as *MEOX1, PDGFRB, SIX1, CRABP2, NR2F1, EYA1,*
FOXC1 and PAX3. PMECs were characterized by EC markers, like *ETV2, CDH5,*
CD34, KDR, ENG, SOX17, PLVAP, APLN, NRP1. The mixed LPM/sclerotome cluster
was characterized by LPM and sclerotome specific genes, such as *TMEM88*, *HAND1,*
TNNI1, PRRX1, ACTA2, DES, FOXH1, LEF1 and *JAG1* (Figure 2G, S3C-D). Diffusion
pseudotime rooted in the paraxial mesoderm increased towards both PMECs and
LPM/Sclerotome (Figure 2H). Both cell types therefore likely differentiated from a
common paraxial mesoderm progenitor.
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225 **Acquisition of an organ-specific identity in cardiac microtissues**
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Being able to produce ECs with properties corresponding to their mesodermal origins
enabled us to test in how far the cellular microenvironment can either reinforce or
reverse this specification i.e. how "plastic" the ECs are. Specifically, we set out to

mimic the cardiac microenvironment *in vitro* using a protocol for creating cardiac MTs,

published previously by our group (Giacomelli et al., 2020). Briefly, CD34+ CMECs or

PMECs were sorted on day 6 and combined with hiPSC-derived cardiomyocytes

(hiPSC-CMs) and hiPSC-derived fibroblasts (hiPSC-CFs) in a ratio of 15:70:15 to form

MTs. MTs made from CMECs (CM_MTs) and PMECs (PM_MTs) were collected after

21 days from two independent biological replicates by scRNA-seq (Figure 3A, Figure

S2A-B). The replicates appeared highly similar in a low-dimensional representation

(Figure S2C) and were therefore combined for further analysis. Any remaining,

undifferentiated hiPSCs were excluded from further analysis (Figure S2F-G). Both

CM_MTs and PM_MTs datasets were divided into three clusters that correspond to

hiPSC-CFs, hiPSC-CMs and hiPSC-ECs (Figure 3B). Marker genes identified for each

cluster, confirmed the cluster identities (Table S6).

In order to assess to what extent ECs in MTs acquired an organ-specific identity, we

compared their expression profiles to primary ECs in a published data set of the

human fetal heart (Asp et al., 2019) (Figure S4A). In this data set, we reannotated the

original endothelium/pericytes/adventia cluster (cluster 10) as intramyocardial ECs,

based on differentially expressed markers such as *A2M, CD36, APLNR, ARHGAP18,*

IGFBP3, CLDN5, FABP4 and *FABP5* (Figure S4B, Table S5). The cluster annotated

as capillary endothelium (cluster 0) in the original publication was reannotated as

endocardium, due to the presence of differentially expressed markers like *NPR3,*

ALDH2, CDH11, ECE1, TMEM100, FOXC1 and *EDN1* (Figure S4B, Table S5).

Supporting the differential expression test, UMAP visualization of representative intra-

myocardial and endocardial markers showed specific expression in the respective

clusters (Figure S4C-D).

We next compared CMECs and PMECs on day 6 of differentiation (CMECs_day 6 and

PMECs_day 6) with the CMECs and PMECs in MTs (CMECs_MT and PMECs_MTs)

respectively. CMECs in MTs upregulated expression of intramyocardial makers, such

as *CLDN5, GMFG, APLNR, CD36, NOTCH4, OIT3, IGFBP3, ARHGAP18, A2M* and

BCAM, but not *FABP5,* compared to CMECs isolated on day 6 of differentiation (Figure

3C,D, Table S7). PMECs in MTs upregulated expression of a few intramyocardial

markers, such as *CLDN5, GMFG, NOTCH4, IGFBP3, ARHGAP18, A2M* and *BCAM,*

but not *APLNR, CD36, OIT3* and *FABP5* compared to PMECs isolated on day 6 of

differentiation (Figure 3C,E, Table S7). We also found some endocardial markers

upregulated in both CMECs and PMECs in MTs (*TFPI2, EDN1, ECE1, FOXP1*)(S5A-

B, Table S7). However, the differences in endocardial marker expression were smaller

compared to intramyocardial markers. Notably, the expression of several

intramyocardial markers, especially *APLNR, CD36, OIT3, ARHGAP18, A2M, BCAM*

and *FABP5* was higher in CMECs in MTs compared to PMECs in MTs (Figure 3F,

Table S8). Although the expression of endocardial markers was also higher in CMECs

in MTs compared to PMECs in MTs, their average expression levels were lower in

general when compared to intramyocardial markers (Figure S5C, Table S8).

Importantly, endocardial makers, including *CDH11, FOXC1, FZD6, TMEM100* and

NPR3, were barely expressed in either CMECs or PMECs in MTs (Figure S5C).

Overall, intramyocardial EC identify was acquired by all hiPSC-ECs in the cardiac MT

environment.

Distinct cell identities are preserved in cardiac microtissues composed of either

cardiac or paraxial mesoderm-derived ECs

To obtain a clearer view of the similarities between ECs in MTs and primary fetal heart

ECs, we merged the CM_MT and PM_MT datasets with the human fetal heart dataset

(Asp et al., 2019) (Figure 4A-B). In the case of both the CM_MT and the PM_MT

dataset, we found that CFs in MTs (CF_MT) clustered together with fetal heart

fibroblast-like cells and CMs in MTs (CM_MT) clustered together with fetal heart

ventricular CMs. Notably, both CMECs in MTs (CMECs_MT) and PMECs in MTs

(PMECs_MT) clustered together with fetal heart intramyocardial ECs and not

endocardium (Figure 4A-D). To quantify our observation, we calculated the distances

(in expression space) between each cell in MTs and the fetal heart dataset. This

calculation showed that CF_MT cells are closest to fibroblast-like cells *in vivo* (related

to cardiac skeleton connective tissue), CM_MT cells are closest to ventricular

cardiomyocytes and CMECs_MT as well as PMECs_MT are closest to intramyocardial

ECs in human fetal heart (Figure 4E). Annotating the *in vitro* cells based on the closest

in vivo neighbors revealed that cell type identities were very similar in CM_MTs and

PM_MTs (Figure 4E).

Correspondingly, the set of markers of the EC cluster in CM_MTs and the PM_MTs, CM_MTs (ECs) and PM_MT (ECs) respectively, showed a high overlap (Jaccard similarity) with the markers of intra-myocardial ECs we extracted from the *in vivo* data set (Figure 4F). The gene set upregulated in CMECs_MT compared to CMECs_day 6 (Up in CMECs_MTs) had a higher overlap with intramyocardial EC markers than the set of genes upregulated in PMECs_MTs compared to PMECs_day 6 (Up in PMECs_MTs) (Figure 4F). For comparison, we also profiled CMECs that were cultured for two additional passages in monoculture. Genes that were upregulated in these cells compared to CMECs_day 6 (Up in P2 CMECs) overlapped the least with intra-myocardial EC markers (Figure 4F). This result excludes the possibility that the effects observed in MTs are simply due to environment-independent differentiation progression over time. CMECs_MT and PMECs_MT thus both resembled intramyocardial ECs but a difference between the two differentiation methods remained. To quantify this difference directly, we used differential gene expression analysis. On day 6 of differentiation, 1446 genes were differentially expressed between CMECs and PMECs (Table S9), while only 332 genes were differentially expressed between CMECs_MT and PMECs_MT (Table S8). 81 genes were shared between the two sets and 251 genes were differentially expressed only between CMECs_MT and PMECs_MT (Figure 4G). Intramyocardial marker genes (*CD36, OIT3, A2M, CLDN5, APLNR, FABP5*) were among these 251 genes that were differentially expressed between the CMECs_MT and PMECs_MT. This is also in line with our observations that expression of some intramyocardial marker genes were higher in CMECs in MTs compared to PMECs in MTs. Next, all EC clusters from bulk and single cell RNA-seq datasets were combined and visualized using principal component analysis (PCA) (Figure 4H). CMECs_day 6 and PMECs_day 6 clustered far apart, while CMECs_MT and PMECs_MT clustered closely together. Bulk and single cell RNA-seq samples clustered together for both CMECs and PMECs. CMECs_MT and PMECs_MT were found closer to fetal heart intramyocardial ECs and fetal heart ECs (Figure 4H) sequenced in our previous study (Giacomelli et al., 2020). Altogether, analysis of the *in vitro* and *in vivo* data sets demonstrated that the cardiac tissue microenvironment resulted in hiPSC-ECs acquiring an intramyocardial EC identity

independent of their developmental origin and, further, gene expression differences

due to distinct mesodermal origins were partially removed.

DISCUSSION

In the present study we derived ECs from hiPSCs from two mesoderm lineages

namely LPM and PM, as previous lineage tracing studies have shown that LPM and

PM serve as a major source of ECs in the developing embryo (Lagha et al., 2009;

Mayeuf-Louchart et al., 2016; Pardanaud et al., 1996). We showed that ECs isolated

on day 6 and day 8 of differentiation retained their developmental lineage history.

Transcription factors involved in the heart and skeletal muscle development were

highly expressed in CMECs and PMECs, respectively. Additional approaches,

including scRNA-seq, showed lineage diversification from a common cardiac and

paraxial mesoderm progenitor during differentiation of CMECs and PMECs,

respectively. At the same time, an organ-specific EC signature was absent in ECs

differentiated either from cardiac or paraxial mesoderm on day 6 and day 8 of

differentiation, or upon extended culture.

Local microenvironmental cues result in acquisition of organ-specific characteristics in

ECs (Aird, 2012). To model the influence of cell-extrinsic factors, we took advantage

of our cardiac MT model which mimics the heart-specific microenvironment, as it

integrates CMs, CFs and ECs (Giacomelli et al., 2020). Although both CMECs and

PMECs acquired an iEC identity after incorporation into MTs, several intramyocardial

EC markers *(APLNR, CD36, OIT3, ARHGAP18, A2M, BCAM* and *FABP5)* were more

strongly upregulated in CMECs compared to PMECs in cardiac MTs. A recent study

in mouse embryos showed that transcriptional heterogeneity in the sinus venosus (SV)

and endocardium-derived iECs declines over time (Phansalkar et al., 2021).

Therefore, it would be interesting to investigate whether PMECs require longer culture

in cardiac MTs to acquire comparable expression of intramyocardial EC markers as in

CMECs. On the other hand, endocardial markers were not detected in either CMECs

or PMECs in cardiac MTs. This is in line with previous evidence that the cardiac MTs

microenvironment recapitulates myocardial and not endocardial layers of the heart.

Importantly, CFs and CMs in both CM_MT and PM_MT datasets were highly similar

and clustered together with fetal heart fibroblast-like cells and fetal heart ventricular

CMs respectively. This showed that developmental origin of ECs does not influence
CF or CM identities in cardiac MTs.

Among EC lineages, PM serves a source of lymphatic ECs in the heart, skin and lymph
node (Lenti et al., 2022; Lupu et al., 2022; Stone and Stainier, 2019). PMECs showed
increased expression of genes important for the development of lymphatic vasculature,
such as *TBX1, FOXC1, LYVE1, VEGFC.* At the same time, expression of the master
regulator of lymphatic EC differentiation (*PROX1*) was not detected in PMECs either
at day 6 or in cardiac MTs. It would be interesting to explore whether addition of known
lymphatic EC growth factors promotes *PROX1* expression in PMECs.

Genetic lineage tracing in mice has identified a variety of developmental origins for
organ-specific ECs in different tissues. However, whether developmental origin is a
prerequisite- or simply a default developmental route remains an open question. On
the other hand, the present and previous studies demonstrate that local
microenvironmental cues might play a bigger role in not only the acquisition but also
the maintenance of the organ-specific characteristics. This is also in line with recent
study on sinusoidal ECs of the liver (Gómez-Salinero et al., 2022). Earlier studies have
demonstrated the importance of the *Gata4* transcription factor in the differentiation of
liver ECs and the acquisition of sinusoidal-like identity (Géraud et al., 2017; Zhang et
al., 2016b). However, scRNA-seq analysis of liver ECs showed that *Gata4* is not
restricted to sinusoidal ECs but it is expressed by all ECs in the liver (Gómez-Salinero
et al., 2022). Instead, the *c-Maf* transcription factor was restricted to sinusoidal liver
ECs and was regulated by BMP9 that is highly expressed by hepatic stellate cells
(HSC) (Breitkopf-Heinlein et al., 2017). Furthermore, overexpression of c-Maf was
sufficient to induce sinusoidal-like characteristics in human umbilical vein ECs
(HUVECs). The same is true for heart ECs that are derived from multiple
developmental pools of EC progenitors that converge to a similar state over time
(Milgrom-Hoffman et al., 2011; Phansalkar et al., 2021; Sharma et al., 2017b).

In summary, we demonstrated that ECs, derived from two distinct mesoderm lineages
namely LPM and PM, acquire organ-specific characteristics upon incorporation into
the cardiac MT environment which includes CMs and CFs. We expect that our findings

will guide the derivation of organ-specific ECs from hiPSCs in the future and lay the

foundation for various biomedical applications, from creating of disease models to

transplantation therapies.

Materials and Methods

hiPSC culture

hiPSC lines LUMC0020iCTRL-06 and NCRM1 (NIH Center for Regenerative Medicine

NIH CRM, obtained from RUDCR Infinite Biologics at Rutgers University) were

cultured in in TeSR-E8 on Vitronectin XF and were routinely passaged once a week

using Gentle Cell Dissociation Reagent (all from Stem Cell Technologies, Vancouver,

Canada).

Generation of PAX3^{Venus}MSGN1^{mCherry} hiPSC dual reporter line

Prior to targeting, NCRM1 hiPSCs were passaged as a bulk on feeders in hESC-

medium (Costa et al., 2007). RevitaCell (Life Technologies, Carlsbad, CA, USA) was

added to the medium (1:200) after every passage to enhance viability after single cell

passaging with TrypLE (Life technologies). PAX3^{Venus} was generated by

CRISPR/Cas9 as follow: NCRM1 hiPSCs were passaged with ratio 1:3 into 60 mm

dishes to reach 60-70% confluency the next day for transfection. Cells were

transfected with pCas9-GFP (Addgene plasmid #44719), pBR322-U6-hPAX3-gRNA-

S1 containing sgRNA CCGGCCAGCGTGGTCATCCT and repair template p15A-cm-

hPAX3-Venus-neo-1kb containing a Venus-neo cassette with 1 kb hPAX3 homology

arms. The antibiotic selection marker is flanked by FRT sites for Flp-mediated

excision. 20 μl lipofectamine (Invitrogen, Waltham, Massachusetts, USA), 8 μg of

pCas9-GFP, 8 μg of sgRNA plasmid and 8 μg of linearized repair template were

diluted in 600 μl of Opti-MEM and added to each 60 mm dish. After 18 hours the

medium was changed to hESC medium. After another 6 hours G-418 (50 μg/ml)

selection was started and was kept for 1 week. Surviving cells were cultured in hESC

medium, passaged and transferred into 6-well plates for the transfection of Flp

recombinase expression vector to remove the neomycin cassette. 300 μl of Opti-MEM

containing 10 μl lipofectamine and 4 μg CAGGs-Flpo-IRES-puro plasmid was added

per well for 18 hours. Puromycin (0,5 μg/ml) selection was started 24 hours post

transfection and lasted for 2 days. Once recovered, cells were passage into 96-well

format for clonal expansion via limited dilution. Targeted clones were identified by PCR
and Sanger sequencing (BaseClear, Leiden, Netherlands). The MSGN1^{mCherry} reporter
line was generated by Transposon mediated BAC transgenesis using protocols
described by (Rostovskaya et al., 2012). In brief, a human BAC (RP11-12L16) with
piggyBac transposon repeats flanking the bacterial backbone and with mCherry
inserted directly after the initiating Methionine of MSGN1 was transfected together with
a piggyBac Transposase into NCRM1 hiPSCs.

Differentiation of ECs from cardiac and paraxial mesoderm

ECs from cardiac mesoderm were differentiated as previously described (Giacomelli
et al., 2017). ECs from paraxial mesoderm were differentiated using a modified Loh et
al. protocol (Loh et al., 2016). Briefly, 5 x 10^4 cells per cm^2 were seeded on plates
coated with 75 µg/mL Matrigel (growth factor reduced) (Corning) the day before
differentiation (day -1). At day 0, paraxial mesoderm was induced by changing TeSR-
E8 to BPEL (Bovine Serum Albumin [BSA], Polyvinyl alcohol, Essential Lipids)
medium (Ng et al., 2008), supplemented with 8 µM CHIR99021. At day 2, cells were
refreshed with BPEL supplemented with 5 µM XAV939. At day 3, cells were refreshed
with BPEL supplemented with 4 µM CHIR99021. From day 5 onwards, cells were
refreshed every 3 days with BPEL medium supplemented with 50 ng/ml VEGF.

Fluorescence-activated cell sorting

448 Cells were dissociated with TrypLE on day 6 and 8 of CMEC and PMEC protocol and
449 stained with VE-Cadherin PE-conjugated Antibody (R&D Systems). Then VEC+ cells
450 were sorted using FACSAria III. Total RNA was extracted right after sorting using the
451 NucleoSpin® RNA kit (Macherey-Nagel, Düren, Germany).

Generation of 3D Cardiac Microtissues (MTs)

Cardiac MTs were generated from hiPSC-derived ECs, CFs and CMs as previously
described (Giacomelli et al., 2020). Briefly, on day 6 of CMEC and PMEC
differentiation, CD34$^+$ ECs were isolated using a Human cord blood CD34 Positive
selection kit II (StemCell Technologies) following the manufacturer's instructions. On
the day of MT formation, freshly isolated hiPSC-ECs and cultured hiPSCs-CFs and
hiPSC-CMs were combined together to 5000 cells (70% cardiomyocytes, 15%

endothelial cells and 15% cardiac fibroblasts) per 50µl BPEL medium supplemented

with VEGF (50 ng/ml) and FGF2 (5 ng/ml). Cell suspensions were seeded on V-bottom

96 well microplates (Greiner bio-one, Kremsmünster, Austria) and centrifuged for 10

min at 1100 rpm. MTs were incubated at 37°C, 5% CO_2 for 21 days with media

refreshed every 3-4 days. scRNAseq analysis of MTs was performed after 21 days.

Fluorescence-activated cell sorting

Cells were detached using TrypLE for 5 min at 37°C and washed once with FACS

buffer (PBS containing 0.5% BSA and 2 mM EDTA). Primary antibodies CD144 (1:50,

eBioscience), CD140a (1:20, BD Bioscience) were added for 1 hr at 4°C. Samples

were measured on MACSQuant VYB (Miltenyi Biotech) equipped with a violet (405

nm), blue (488 nm) and yellow (561 nm) laser. The results were analyzed using Flowjo

v10 (FlowJo, LLC). For FACS sorting on day 6 and day 8 of the CMEC and PMEC

protocol, CD144+CD140a- cells were sorted using FACSAria III (BD-Biosciences).

Quantitative Real-Time Polymerase Chain Reaction (qPCR)

Total RNA was extracted using the NucleoSpin® RNA kit according to the

manufacturer's protocol. cDNA was synthesized using an iScript-cDNA Synthesis kit

(Bio-Rad, Hercules, CA, USA). iTaq Universal SYBR Green Supermixes (Bio-Rad)

and Bio-Rad CFX384 real-time system were used for the PCR reaction and detection.

Relative gene expression was determined according to the standard ΔCT calculation

and normalized to the housekeeping gene RPL37A.

Bulk RNA sequencing and analysis

Bulk RNAseq of passage two CMECs (CMECs P2) and human fetal heart ECs at

gestation age Week(W)12.5, W15 and W21 were performed in our previous study

(Giacomelli et al., 2020) and obtained from GEO accession number GSE116464.

Bulk RNAseq of day 6 and 8 of CMEC and PMEC differentiation were performed at

BGI (Shenzhen, China) using the Illumina Hiseq4000 (100bp paired end reads). Raw

data   was   processed   using   the   LUMC   BIOPET   Gentrap   pipeline

(https://github.com/biopet/biopet), which comprises FASTQ preprocessing, alignment

and read quantification. Sickle (v1.2) was used to trim low-quality read ends

(https://github.com/najoshi/sickle). Cutadapt (v1.1) was used for adapter clipping

(Martin, 2011), reads were aligned to the human reference genome GRCh38 using

GSNAP (gmap-2014-12-23) (Wu and Nacu, 2010; Wu and Watanabe, 2005) and gene

read quantification with htseq-count (v0.6.1p1) against the Ensembl v87 annotation

(Yates et al., 2015). Gene length and GC content bias were normalized using the R

package cqn (v1.28.1) (Hansen et al., 2012).Genes were excluded if the number of

reads was below 5 reads in ≥90% of the samples.

Differentially expressed genes were identified using generalized linear models as

implemented in edgeR (3.24.3) (Robinson et al., 2009). P-values were adjusted using

the Benjamini-Hochberg procedure and FDR ≤ 0.05 was considered significant.

Analyses were performed using R (version 3.5.2). PCA plot was generated with the

built-in R functions prcomp using transposed normalized RPKM matrix. Correlation

among samples was calculated using cor function with spearman method and the

correlation heatmap was generated with aheatmap function (NMF package). Gene

ontology enrichment was performed using compareCluster function of clusterProfiler

package (v3.10.1) (Yu et al., 2012) and q ≤ 0.05 was considered significant.

Single-cell RNA sequencing and analysis

Library preparation and sequencing

Library preparation was performed as previously described (Giacomelli et al., 2020).

Briefly, single cells were loaded into the 10X Chromium Controller for library

construction using the Single-Cell 3' Library Kit, Version 2 Chemistry (10x Genomics,

Pleasanton, CA, USA) according to the manufacturer's protocol. Next, indexed cDNA

libraries were sequenced on the HiSeq4000 platform. Single-cell expression was

quantified using unique molecular identifiers (UMIs) by 10x Genomics' "Cell Ranger"

software.

The mean reads per cell for all eight data sets: CMEC (R1): 28,499; CMEC (R2):

29,388; PMEC (R1): 31,860; PMEC (R2): 38,415; CM_MT (R1): 39,319; CM_MT (R2):

29,741; PM_MT (R1): 36,726; PM_MT (R2): 26,421.

Single-cell RNAseq data pruning and normalization

For data pruning and normalization, the two replicates of each of the 4 conditions

(CMEC, PMEC, CM_MT and PM_MT) were combined without batch correction. Then,

cells with a low number of genes per cell (1200 [CMEC], 1200 [PMEC], 900 [CM_MT],

750 [PM_MT], see Fig. S2A-B) were removed. Genes expressed in less than 2 of the

remaining cells with a count of at most 1 were excluded from further analysis. Each

combined data set was normalized using the R package scran (V 1.14.6) (Lun et al.,

2016). Highly variable genes (HVGs) were calculated (improvedCV2 from the scran

package) for each replicate of the combined data sets after excluding ribosomal genes

[Ribosomal Protein Gene Database], stress markers (Brink et al., 2017) and

mitochondrial genes. For downstream analysis the top 5% HVGs were used after

excluding proliferation (Whitfield et al., 2006) and cell cycle (Giotti et al., 2017) related

genes.

Cell cycle analysis and batch correction

For each combined data set, cell cycle analysis was performed with the scran package

using the cyclone function (Scialdone et al., 2015) on normalized counts (Figure S2H).

Cells with a G2/M score higher than 0.2 were considered to be in G2/M phase.

Otherwise, they were classified as G1/S. Using this binary classifier as predictor, we

regressed out cell cycle effects with the R package limma (V 3.42.2) (Ritchie et al.,

2015) applied to log-transformed normalized counts. Then, for each combined data

set, the two replicates were batch corrected with fast mutual nearest neighbors

correction method (MNN) (Haghverdi et al., 2016) on the cell cycle corrected counts,

using the 30 first principal components and 20 nearest-neighbors (Figure S2C).

Clustering

For each combined data set, batch-corrected counts were standardized per gene and

then used to create a shared nearest neighbour (SNN) graph with the *scran* R package

(d = 30, k = 2). Louvain clustering was applied to the SNN graph using the *igraph*

python package (V 0.7.1) with these resolution parameters: 0.4 [CMEC], 0.4 [CM_MT],

0.3 [PMEC], 0.1 [PM_MT]. For the CMEC data set, this resulted in 5 clusters (Figure

S2D). Two of these 5 clusters were excluded from further analysis based on the

expression of pluripotency markers (Figure S2E). For the PMEC data set, this resulted

in 3 clusters (Figure 2F). For CM_MT and PM_MT, clustering resulted in 4 clusters

(Figure S2F and S2G), where one cluster was excluded from further analysis, because
it was mainly present in one of the two replicates. Additionally, the attempt to map this
cluster to in vivo data, resulted in mostly unassigned cell types (plot not shown). For
PMEC, clustering resulted in 3 clusters.
562
Dimensionality reduction and pseudotime
Dimensionality reduction was performed using the python *scanpy* pipeline (V 1.4.6).
For both data sets, CMEC and PMEC, a 20 nearest-neighbors (knn, k=20) graph was
created from diffusion components of the batch corrected data sets. Diffusion
components are the eigenvectors of the diffusion operator which is calculated from
Euclidean distances and a gaussian kernel. The aim is to find a lower dimensional
embedding which considers the cellular progression. Both graphs were projected into
two dimensions with the default force-directed graph layout and starting positions
obtained from the partition-based graph abstraction (PAGA) output (Wolf et al., 2019).
PAGA estimates connectivities between partitions and performs an improved version
of diffusion pseudotime. Diffusion pseudotime (Haghverdi et al., 2016; Wolf et al.,
2019) was calculated on these graphs with root cells selected based on the graph
layout from the "Cardiac Mesoderm" cluster in CMEC, and the "Paraxial Mesoderm"
cluster in PMEC.
577
578   For CM_MT and PM_MT, the knn graphs (k=50 for PM_MT, k=100 for CM_MT) were
579   created from the first 30 principal components of the batch corrected data sets. These
580   graphs were projected into two dimensions with the default force-directed graph layout
581   and starting positions from the PAGA output.
582
***In vivo* data analysis and mapping**
The *in vivo* data set, downloaded from https://www.spatialresearch.org/resources-
published-datasets/doi-10-1016-j-cell-2019-11-025/, contains a 6.5 PCW human
embryonic cardiac tissue sample. The clusters and cluster annotations were obtained
from the original publication (Asp et al., 2019). The data set was normalized with the
scran R package and HVGs were calculated as described in section "Single-cell RNA-
seq data pruning and normalization". Dimensionality reduction was performed with the

R package *umap* (V 0.2.5.0) using 20 nearest-neighbors, min_dist = 0.7 and Euclidean
distance.

Differential expression analysis
All differential expression tests were performed with edgeR (V 3.28.1) (Robinson et
al., 2009) using a negative binomial regression and raw counts. The predictors in the
regression were: cluster and replicate (both discrete variables), as well as the total
number of counts per cell.

For marker gene analysis (Figures S3A and S3C), p-values were obtained for a
contrast between the cluster of interest and all other clusters using regression
coefficients averaged over the replicates. For tests between different data sets (Figure
3C), the corresponding endothelial cell cluster was extracted from each data set. Then,
a contrast between MT and day 6 was calculated by averaging over the predictors of
both replicates. For the *in vivo* test (Figure S4B), intra-myocardial EC and
endocardium clusters were extracted from the data set to calculate the contrast
between them. P-values were adjusted for multiple hypothesis testing with the
Benjamini-Hochberg method.

Comparison to the *in vivo* data set
CM_MT and PM_MT data sets were mapped on the *in vivo* data set using the MNN
method (d = 30 principal components, k = 100 nearest neighbors). First, *in vitro*
replicates were mapped to each other, then the *in vivo* data was mapped on the
combined *in vitro* data, using normalized, log- transformed counts and the 10% top
HVGs of the *in vivo* data set. Dimensionality reduction was performed with the R
package *umap* using 100 nearest-neighbors, min_dist = 0.3 and Euclidean distance.
K-nearest-neighbour (KNN) assignment was performed in the batch corrected,
principal component space (30 PCs). The 100 nearest-neighbors in the *in vivo* data
set based on Euclidean distances were calculated for each *in vitro* cell. The *in vitro*
cell was ascribed the cell type most abundant among the 100 *in vivo* neighbors. Each
such assignment received a confidence score, which is the number of *in vivo*
neighbors with that cell type divided by the number of all nearest neighbors (=100). A
cell was not ascribed a cell type if either the average distance to its nearest neighbour

exceeded a certain threshold (determined by the long tail of the histogram of average

distances: 0.35), or the assignment had a confidence score less than 0.5. In addition,

clusters containing less than 10 cells were not ascribed a cell type.

For the Jaccard similarity measure, marker genes of each differential expression test

were selected with adjusted p-value < 0.05. The remaining genes were ranked by log2

fold change and the first 478 genes were selected for analysis. Then, the Jaccard

distances were calculated between the marker genes of intra-myocardial endothelial

cells and each of the other gene sets.

For principal component analysis (Figure 4H), human fetal bulk samples and in vitro

bulk samples (Giacomelli et al., 2020) were combined with the single cell data sets.

For each single-cell data set, the endothelial cells were extracted and the sum per

gene over all cells was calculated. Then, bulk and single cell samples were log-

transformed and combined into one data set. Principal component analysis was

applied on the gene-wise standardized data set, using marker genes of the intra-

myocardial endothelial cells from the *in vivo* data set.

Data Availability Statement

The accession numbers for the bulk and single cell RNA sequencing datasets reported

in this paper are https://www.ncbi.nlm.nih. gov/geo/ GEO: GSE151427 (day 6 and day

8 CMEC and PMEC (bulk); day 6 PMEC (single cell); PM_MT (single cell));

GSE202901 (day 6 CMEC (single cell)); GSE147694 (CM_MT (single cell)).

Statistics

Statistical analysis was conducted with GraphPad Prism 7 software (San Diego, CA,

USA). Data are represented as mean± SD.

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Author contributions

X.C., M.M., S.S., V.V.O. designed the research, analyzed and interpreted results,

wrote the manuscript; X.C., F.E.vdH, V.V.O. performed experiments; M.M., S.S.

performed scRNA-Seq analysis; K.N. generated MSGN1^{mCherry} hiPSC reporter line;

K.A. generated gene targeting constructs for reporter hiPSC lines; X.C., H.M.

performed bulk RNA-seq analysis; C.L.M. designed the research and edited the

manuscript.

Competing interests

673   C.L.M. is co-founder of Ncardia bv. The other authors indicated no potential conflicts

674   of interest.

Figure Legends

Figure 1. Characterization of ECs differentiated from hiPSCs using CMEC and PMEC protocols

(**A-B**) Schematic overview of CMEC (**A**) and PMEC (**B**) differentiation protocols.

Endothelial cell (CD144+CD140a-) were FACS sorted on days 6 and day 8 for bulk

RNA-seq. ACTA: activin-A. CHIR: CHIR99021. pPSM/aPSM: posterior/anterior

presomitic mesoderm. LPM: lateral plate mesoderm. (**C**) Flow cytometry analysis of

CD140a and CD144 expression on day 6 and day 8 of CMEC and PMEC

differentiation. (**D**) Quantification of VEC (CD144)+CD140a- cells on day 6 and day 8

of CMEC and PMEC differentiation. Error bars represented standard deviations

calculated from five to six independent differentiations. (**E**) PCA analysis of ECs sorted

on day 6 and 8 of the CMEC or PMEC protocol. (**F**) GO term enrichment analysis for

DEGs between CMECs and PMECs on day 6 of differentiation. Complete list of GO

terms can be found in Table S2. Color represents the $p_{adjusted}$ of enrichment analysis

and dot size represents the count of genes mapped to the GO term. (**G**) Normalized

gene expression levels (RPKM) of cardiac and skeletal related genes in CMECs and

PMECs on day 6 and 8.

Figure 2. Single-cell RNA sequencing analysis of ECs differentiated from

cardiac and paraxial mesoderm

(**A-B**) Schematic overview of CMEC (**A**) and PMEC (**B**) differentiation protocols until

day 6. Cells were collected for scRNA-seq on day 6. ACTA: activin-A. CHIR:

CHIR99021. pPSM/aPSM: posterior/anterior presomitic mesoderm. LPM: lateral plate

mesoderm. (**C**) scRNA-seq of CMEC differentiation on day 6 (PAGA plot). Three

clusters of cells, indicated by color, were identified. (**D**) PAGA plots show expression

of *MESP1, ISL1, SMARCD3, CDH5, MYL4* on day 6 of CMEC differentiation. Color

represents log transformed expression. (**E**) Diffusion pseudotime analysis of CMEC

differentiation on day 6. (**F**) scRNA-seq of PMEC differentiation on day 6 (PAGA plot).

Three clusters of cells, indicated by color, were identified. (**G**) PAGA plots show

expression *PAX3, EYA1, FOXC1, CDH5, PPRX1* on day 6 of CMEC differentiation.

Color represents log transformed expression. (**H**) Diffusion pseudotime analysis of

PMEC differentiation on day 6.

Figure 3. hiPSC-ECs acquired organ-specific signatures in a cardiac

microenvironment

(**A**) Generation of cardiac MTs from hiPSC-CMs, hiPSC-CFs and hiPSC-ECs. CMECs

and PMECs were used for CM_MTs and PM_MTs respectively. MTs were collected

after 21 days for scRNAseq. (**B**) scRNAseq data of CM_MTs (left) and PM_MTs (right)

were visualized using PAGA. Three clusters of cells were identified in both datasets.

(**C**) Volcano plot showing fold changes and p-values of differential expression tests

between CMECs_day 6 and CMECs_MTs (left), or PMECs_day 6 and PMECs_MTs

(right). Representative intra-myocardial and endocardial markers that are differentially

expressed ($p_{adjusted}<0.05$) are highlighted in red and green, respectively. (**D-F**) Violin

plots of gene expression in PMECs_ day 6, CMECs_day6, CMECs_MT and

PMECs_MT for representative Intra-myocardial EC markers. Asterisks indicate

significance level of differential gene expression tests between the respective

populations. ns: $p>0.05$; * $p <= 0.05$; ** $p<= 1e-10$; *** $p<= 1e-100$; **** $p<= 1e-200$.

The clusters with higher expression value were labeled. ne: not expressed (0 counts)

in >85% of cells in both groups.

Figure 4. CMECs and PMECs acquired intramyocardial identity in cardiac

microenvironment

(**A-B**) Low-dimensional representation (UMAP) of the human fetal heart data set (Asp

et al., 2019) integrated with the CM_MTs (**A, C**) or PM_MTs (**B, D**) dataset. EC clusters

are marked with squares. Cell clusters are indicated by color. Cells in fetal heart

dataset are presented with dots and cells in MTs datasets are presented with contour

lines. (**C-D**) Zoom-ins of the EC clusters marked in (**A-B**). (**E**) Annotation of CM_MT

and PM_MT cells based on nearest neighbors in the *in vivo* dataset. Cells from

CM_MTs and PM_MTs are outlined with a green or violet contour, respectively. (**F**)

Jaccard similarity to intramyocardial ECs from the human fetal heart dataset was

calculated for each indicated group of genes. CM_MTs (ECs): specific markers of

cluster CMECs within the CM_MT dataset; PM_MTs (ECs): specific markers of cluster

PMECs within the PM_MT dataset; Up in CMECs_MT: DEGs that are higher in

CMECs_MT compared to CMECs_day 6; Up in PMECs_MT: DEGs that are higher in

PMECs_MT compared to PMECs_day 6; Up in P2 CMECs: DEGs that are higher in

passage two CMECs compared to CMECs_day 6. (**G**) Venn diagram shows numbers

and overlap of DEGs ($p_{adjusted}<0.05$ and fold-change > 1.5) between CMECs and

PMECs from day 6 (in red) and MTs (in blue). (**H**) PCA plot of different EC populations

in scRNAseq (triangle and diamond) and bulk RNA-seq (circle) datasets. Average

expression values of all cells in the cluster were used for the scRNA-seq data.

Figure S1. Characterization of PMEC differentiation using

MSGN1^{mCherry}PAX3^{Venus} dual reporter line

(**A**) Targeting constructs used to generate PAX3-Venus and MSGN1-mCherry hiPSC

reporter line. (**B**) Flow cytometry analysis of PAX3^{Venus} and MSGN1^{mCherry} expression

on day 2, 3, 5, 6 and 8 of PMEC differentiation. (**C-D**) Quantification of mCherry+ (**C**)

and Venus+ (**D**) cells in the total population by flow cytometry on day 2, 3, 5, 6 and 8.

(**E**) Representative fluorescence image of PAX3^{Venus} expression on day 5 of PMEC

differentiation. Scale bar: 200 μm. (**F**) Quantification of *TBXT, MIXL1, MESP1,*

GATA4, NKX2-5, MSGN1, TBX6 and *PAX3* expression by qPCR on day 0, 2, 5 and 8

of CMEC (green) and PMEC (purple) differentiation.

Figure S2. Quality control of scRNAseq datasets

(**A-B**) Distribution of the number of detected genes (**A**) and total expression (**B**) in

each cell of the scRNAseq datasets. The dotted blue lines indicate quality control

thresholds. Two different batches are labelled with different colors. (**C**) Two different

batches of cells collected for each scRNAseq dataset were visualized with PAGA.

(**D**) scRNAseq data of CMECs on day 6 is visualized using PAGA. Five cell clusters

were identified and labelled with different colors. (**E**) Expression of pluripotency genes

POU5F1 and *NANOG* in the CMECa dataset on day 6 is shown in PAGA plot. Color

represents log transformed expression. (**F-G**) scRNAseq data of CM_MTs (**F**) and

PM_MTs (**G**) were visualized using PAGA. Four cell clusters were identified. Clusters

labelled with "Other" were excluded from downstream analysis. (**H**) Boxplot of G2M

phase-score in individual clusters of each dataset.

Figure S3. scRNAseq analysis of CMEC and PMEC datasets on day 6

(**A**) Volcano plots showing fold changes and p-values of differential expression tests

between individual CMEC clusters and all other cells. Representative significantly up-

regulated genes ($p_{adjusted}$ < 0.05 & fold-change > 1.2) are labelled in red. (**B**) *MESP1,*

SMARCD3, *KDR*, *CDH5*, *MYL4* and *NKX2-5* expression (log transformed) in three

clusters of the CMEC dataset on day 6. (**C**) Volcano plots showing fold-changes and

p-values of differential expression tests between individual PMEC clusters and all

other cells. Representative significantly up-regulated genes ($p_{adjusted}$<0.05 & fold

change > 1.2) are labelled in red. (**D**) *PAX3*, *MEOX1*, *CD34*, *CDH5*, *HAND1* and

PRRX1 expression (log transformed) in three clusters of PMEC dataset on day 6.

Figure S4. Re-analysis of a published scRNAseq dataset to identify organ

specific signatures of human fetal heart ECs

(**A**) Low-dimensional representation (UMAP) of scRNAseq of the human fetal heart

(Asp et al., 2019). 14 cell clusters were identified and named based on the original

publication, except for two EC clusters: intramyocardial ECs and endocardium. (**B**)

Volcano plot showing fold changes and p-values for differential expression tests

between intra-myocardial ECs and endocardium in the data set shown in (**A**).

Representative differentially expressed genes ($p_{adjusted}<0.05$) that are known as

intramyocardial and endocardial markers are labelled in red and green respectively.

(**C-D**) Low-dimensional representation (UMAP) of scRNAseq of the human fetal heart

(Asp et al., 2019). Log-transformed expression of representative intramyocardial EC

markers (**C**) and endocardium markers (**D**) is indicated by color.

Figure S5. Comparison of organ-specific signatures of hiPSC-ECs on day 6 with

ECs in MTs

(**A-C**) Differential expression test between clusters CMECs_day 6 and CMECs_MT

(**A**), PMECs_ day 6 and PMECs_MT (**B**), CMECs_MT and PMECs_MT (**C**) for

representative endocardial EC markers. ns: $p>0.05$; * $p <=0.05$; ** $p<= 1e-10$; *** $p<=$

1e-100; **** $p<= 1e-200$. Clusters with higher expression value were indicated with

stars. ne: not expressed (0 counts) in >85% of cells in both groups.

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Figure 1.

Figure 2.

Figure 3.

Figure 4.

Figure S1.

Figure S2.

Figure S3.

Figure S4.

Figure S5.