INTRODUCTION

Diabetes mellitus is a metabolic syndrome characterized by elevated blood glucose levels that result from reductions in insulin production or action. Insulin is produced by pancreatic β cells found within the endocrine islets of Langerhans. A potential treatment for diabetes is to replace insulin by transplantation of human embryonic stem cell (hESC)-derived β cells. Derivation of functional β cells from hESCs requires an in-depth understanding of how endocrine cells form during embryonic development.

During mouse and human pancreas development, pancreatic progenitors become restricted to the endocrine cell fate before differentiating to hormone-producing cells. This process involves many transcription factors (TFs) that drive the changes in gene expression necessary for endocrine cell genesis. Genetic loss-of-function studies have found roles for individual TFs in the formation of specific islet cell types. From this work, a map of the TF cascade that regulates the formation of endocrine cells, including the β cells, has emerged (Cano et al., 2014). However, our understanding of fate decisions during endocrine cell formation is based on studies of the whole population of progenitors, using technologies such as bulk RNA sequencing and often only in mouse cells. The gene expression of individual human and mouse cells during terminal differentiation is unknown.

A promising method to understand gene expression changes at single-cell resolution is single-cell RNA sequencing (scRNA-seq). Following the first publication in 2009 (Tang et al., 2009), commercial platforms and lower sequencing costs have made scRNA-seq a feasible technology for many biologists. Recently, several studies have investigated the single-cell transcriptome of healthy and type 2 diabetes human islets (Baron et al., 2016; Chu et al., 2016; Enge et al., 2017; Lawlor et al., 2017; Qiu et al., 2017; Segerstolpe et al., 2016; Wang et al., 2016; Xin et al., 2016; Zeng et al., 2017). From these studies, we have begun to appreciate the cell-type-specific gene expression changes that occur during diabetes progression, the differences between mouse and human islet cells, and the identity of islet and pancreatic cell types.

Several recent studies have started characterizing the single-cell transcriptomes of mouse and human progenitors during embryonic development, including mouse embryonic day (E) 13.5 (Stanescu et al., 2017), E14.5 (Byrnes et al., 2018), E16.5 (Scavuzzo et al., 2018), and E17.5 pancreas (Byrnes et al., 2018) and human fetal pancreas (Ramond et al., 2018). In this study, scRNA-seq was used to analyze 6,905 E15.5 pancreatic cells, 6,626 E18.5 pancreatic cells, and 4,462 hESC-derived endocrine progenitor (EP) cells. From these data, novel cell types were identified and comparisons between hESC-derived endocrine cells, mouse EP, and human islet cells were made. Characterization of these populations will aid efforts to generate an...
Figure 1. Cell Populations in E15.5 and E18.5 Mouse Pancreas

(A) Schematic overview of the two transgenic mouse lines used to isolate cell populations during pancreas development. Using this strategy, pancreatic progenitors (P; red) are tdTomato+, early Neurog3-lineage cells (N\text{Early}; yellow) are tdTomato+ and eGFP+, and late Neurog3-lineage cells (N\text{Late}; green) are eGFP+.

(B and C) FACS plot of (B) E15.5 and (C) E18.5 cells used for library generation.

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unlimited source of insulin-producing β-like cells for diabetes treatment.

RESULTS

Strategy for Generating Single-Cell Transcriptomes of Embryonic Mouse Pancreas

Two transgenic mouse lines were used to isolate progenitor populations during pancreas development: Neurog3-Cre and Rosa26<sup>mtm<sup>G</sup> (Figure 1A). In Neurog3-Cre; Rosa26<sup>mtm<sup>G</sup></sup> embryos, all cells are labeled with a membrane-targeted Tomato red fluorescent protein (tdTomato). Upon activation of the Neurog3 promoter, Cre recombinase removes the floxed tdTomato cassette, resulting in expression of a membrane-targeted enhanced green fluorescent protein (eGFP). Cells that recently activated Neurog3 express both tdTomato and eGFP (yellow; N<sup>Late</sup>), while cells that are further along along the endocrine cell lineage express eGFP only (green; N<sup>Late</sup>) (Figure 1A) (Xu et al., 2015). This strategy was used to isolate by fluorescence-activated cell sorting (FACS) the three populations from the pancreas of one E15.5 and E18.5 embryo, by fluorescence-activated cell sorting (FACS) the three populations during pancreas development: Neurog3-Cre and Rosa26<sup>mtm<sup>G</sup></sup> embryos, all cells are labeled with a membrane-targeted Tomato red fluorescent protein (tdTomato). Upon activation of the Neurog3 promoter, Cre recombinase removes the floxed tdTomato cassette, resulting in expression of a membrane-targeted enhanced green fluorescent protein (eGFP). Cells that recently activated Neurog3 express both tdTomato and eGFP (yellow; N<sup>Late</sup>), while cells that are further along along the endocrine cell lineage express eGFP only (green; N<sup>Late</sup>) (Figure 1A) (Xu et al., 2015). This strategy was used to isolate by fluorescence-activated cell sorting (FACS) the three populations from the pancreas of one E15.5 and E18.5 embryo, and single-cell libraries were generated using a 10<sup>x</sup> Genomics Chromium single cell 3’ kit. In total, 7,502 E15.5 and 7,023 E18.5 single cells were sequenced at a depth of >50,000 reads per cell using Illumina NextSeq 500 (Table 1).

Identification of Cell Types in E15.5 and E18.5 Pancreas

At E15.5, 91% of pancreatic cells expressed tdTomato protein only (Figure 1B). As the yellow and green populations were less abundant (6.8%), these cells were pooled together and sequenced as one library (Figure 1B). At E18.5, the red (53%), yellow (1.5%), and green (2.6%) cells were used to generate three separate libraries (Figure 1C). To explore the cell types present within the pancreas, the sequenced red, yellow, and green cells at E15.5 and E18.5 were aggregated into single datasets using cellranger aggr and low-quality cells were excluded from analysis using Scatter and Seurat (see Experimental Procedures for details). Following this, 6,905 E15.5 and 6,626 E18.5 cells were clustered using unsupervised k-means clustering and visualized using t-distributed stochastic neighbor embedding (t-SNE) (van der Maaten and Hinton, 2008). Cluster identity was inferred using differentially expressed genes (Tables S2 and S3 and Figure S1). At 15.5, there were 15 clusters representing mesenchyme cells (50.0%), acinar cells (8.8%), bipotent trunk progenitor cells (7.9%), ductal cells (6.6%), Chga-expressing immature endocrine cells (6.3%), α cells (6.3%), EPs (4.7%), macrophages (3.1%), β cells (2.6%), neuronal cells (1.9%), and endothelial cells (1.9%) (Figure 1D). Many of these clusters were also present at 18.5, including mesenchyme cells (22.8%), α cells (17.2%), β cells (14.6%), trunk cells (9.3%), endothelial cells (6.0%), acinar (3.1%), ductal (3.0%), macrophages (2.7%), and neuronal cells (1.1%) (Figure 1E). As we sequenced more yellow and green cells at E18.5 (3,950 versus 1,322 cells) (Table S1), three other endocrine populations were found: δ cells (9.4%), immature endocrine cells (Endo; 9.4%), and cells undergoing DNA replication (S-phase; 1.3%) (Figure 1E).

To verify the library identity of individual cells, the GRCm38 genome used for alignment was annotated to include the sequences for the tdTomato and eGFP transgenes. At E15.5 and E18.5, the trunk, acinar, ductal, mesenchymal, endothelial, neuronal, and macrophage cells expressed tdTomato, consistent with the non-endocrine lineage of these cell types (Figures 1F and 1G). At E15.5, the EP cells co-expressed both tdTomato and eGFP (Figure 1F), suggesting recent activation of Neurog3. In addition, a subset of Endo cells at E18.5 also co-expressed tdTomato and eGFP, consistent with a small EP population (Figure 1G). Interestingly, a subset of the trunk cells expressed eGFP at both E15.5 and E18.5 (Figures 1F and 1G), consistent with heterogeneous activation of Neurog3 transcription in trunk cells. As all endocrine cells are derived from Neurog3<sup>+</sup> progenitors (Gu et al., 2002), E15.5 and E18.5 endocrine cells expressed eGFP (Figures 1F and 1G). Taken together, scRNA-seq identified pancreatic cell populations and differentially expressed genes at E15.5 and E18.5.

Characterization of the Mouse Embryonic Endocrine Cell Transcriptome

To understand the transcriptional changes that occur during endocrine specification, the yellow and green cells were further characterized at E15.5. After filtering, 1,322 cells were analyzed using unsupervised k-means clustering and visualized using a t-SNE plot (Figure 2A). Eight clusters representing several cell populations were identified using the top ten differentially expressed genes (Table S4): α cells (20.3%), two EP populations (EP1, 19.1%, and EP2, 10.5%), β cells (18.8%), Chga-expressing endocrine cells (15.0%), δ cells (9.4%), immature endocrine cells (Endo; 9.4%), and cells undergoing DNA replication (S-phase; 1.3%).
Endo cells (7.3%), trunk progenitor cells (4.8%), and ghrelin cells (6.7%).

As Neurog3+ progenitor cells exit the cell cycle during differentiation to endocrine cells (Desgraz and Herrera, 2009; Jensen et al., 2000; Miyatsuka et al., 2011), the cell cycle stage of individual cells at E15.5 was investigated. While the EP clusters included dividing cells, cells of the endocrine lineage mainly expressed G0/1 markers, consistent with cell cycle exit (Figure S2A). Interestingly, the trunk population also contained many S- and G2/M-phase cells that expressed tdTomato, eGFP, and Neurog3, suggesting that these trunk cells recently activated the Neurog3 promoter (Figures S2A–S2C) (Bechard et al., 2016).

To confirm the identity of the E15.5 yellow and green clusters, the expression of several genes was investigated. Neurog3 was highly expressed in both EP clusters, while lower expression was also found in trunk, ε cells, and Chg-expressing cells (Figures 2B and S2C). Neurod1, a target of Neurog3, was expressed throughout the endocrine cell lineage except in the trunk cells (Figure S2C). Both Ins1 and Ins2 were expressed in the β cells while Gcg and Ghrl were specific to the α and ε cells, respectively (Figures 2C–2F and S2C).

To find cell-type-specific markers, the top ten differentially expressed genes in the EP1 (yellow), EP2 (blue), α (red), and β (green) cells were profiled (Figure S2D). The EP clusters expressed known marker genes such as Neurog3, Pax4, and Sox4 along with previously undescribed genes, including Midkine (Midk) and Growth Arrest and DNA Damage Inducible Alpha (Gadd45a) (Figure S2D). The α cells were enriched for expression of Gcg and previously proposed α cell markers Sfk38a5 (Stanescu et al., 2017) and Transhyretin (Ttr) (Westmark and Westmark, 2008). The β cell cluster expressed several β cell genes, including Ins1, Ins2, Pdx1, and Iapp (Figure S2D), along with the Neurod1 target Nnat (Chu and Tsai, 2005) and previously identified β cell marker Ppp1r1a (Martens et al., 2011).

We next aimed to characterize cells of the endocrine lineage at E18.5. To do this, cells from the E18 yellow and E18 green libraries were aggregated using cellranger aggr, resulting in 3,970 cells made up of 561 yellow and 3,409 green cells (Table S1). Visualizing these data using t-SNE revealed 11 clusters: two β cell populations (B1, 20%, and B2, 11.1%), α cells (21.0%), δ cells (15.0%), trunk cells (10.1%), ε cells (7.7%), Chg-expressing Endo cells (6.1%), S-phase cells (3.8%), stellate cells (2.7%), EPs (1.8%), and macrophages (0.6%) (Figure 2G). Yellow library cells were found in the trunk, EP, stellate, and macrophage cells and expressed tdTomato and eGFP (Figures S3A and S3B). Many of the same genes (Spp1, Mt1, and Mt2) were expressed in trunk cells at E15.5 and E18.5, while EP cells expressed Neurog3, Gadd45a, and Pax4 (Figures S2E and S3C).

The endocrine cells were found within the E18.5 green cell library and expressed eGFP only (Figures S3A and S3B). Both Ins1 and Ins2 were highly expressed in the two β cell populations, B1 and B2, as well as in the S-phase cells (Figures 2H, 2I, and S3C), consistent with the start of the wave of replication that is required for the acquisition of β cell mass (Georgia and Bhushan, 2004). Differential expression analyses of the β cell populations reveal cluster-specific differences in gene expression (Table S5). The top ten differentially expressed genes in the β1 cluster included markers of mature cells, including Ins1, Ins2, G6pc2, and Scl2a2, while the β2 cluster was a Nnat-expressing immature cell state (Figure S3E). The Chg-expressing Endo cell state also expressed Nnat along with the progenitor cell marker Neurod1 (Figure S3E). The S-phase cells expressed high levels of Ins1, Ins2, and Gcg, suggesting this cluster represents a mixture of α and β cells (Figure S3C) undergoing DNA replication and mitosis (Figure S3E). As expected, Sst expression was upregulated in the δ cell population (Figure 2J), Ghrl was upregulated in the ε cell cluster (Figure 2L), and Gcg was highly expressed in the α cell cluster (Figure 2M). Although none of the clusters showed specific upregulation of Ppy at this stage, some α cells and δ cells expressed Ppy (Figure 2K).

**Pseudotime Analysis of Endocrine Lineage Using RNA Velocity**

To understand the developmental cell lineage and resultant changes in gene expression, RNA velocity was used to order the E15.5 and E18.5 endocrine cells in pseudotime (Manno et al., 2018). RNA velocity uses unspliced and spliced mRNA to estimate the rate of change between cell states. Using the E15.5 aggr library (from Figure 1), a reduced dataset of 2,869 cells was generated using the trunk, acinar, ductal, EP, Chg, α, and β cell populations and visualized using a t-SNE plot (Figure 3A). Based on the ratio of unspliced/spliced transcripts (Figure S4), a velocity was assigned, and the average vector field was represented by an arrow indicating the direction of differentiation. Within the ductal
Figure 2. Cell Populations in Mouse Endocrine-Lineage Cells at E15.5 and E18.5

(A) t-SNE plot of eight clusters in E15.5 yellow and green cells. These clusters were identified as trunk cells (Trunk; 4.8%), two endocrine progenitor clusters (EP1 yellow, 19.1%, and EP2 blue, 10.5%), Chga-expressing immature endocrine cells (Chg; 15.0%), endocrine cells (Endo; 7.3%), α cells (α; 20.3%), β cells (β; 15.0%), and ghrelin cells (γ; 4.3%).

(B–F) Single-cell gene expression of (B) Neurog3, (C) Ins1, (D) Ins2, (E) Gcg, and (F) Ghrl across cell clusters.

(G) t-SNE plot of 11 cell clusters from E18.5 yellow and green cells: trunk (10.1%), endocrine progenitor (EP; 1.8%), Chg-expressing (Chg; 6.1%), α cell (α; 21.0%), two β cell (β1, 20%, and β2, 11.2%), δ cell (δ; 15.0%), Ghrl cell (7.7%), S-phase cell (S; 3.8%), stellate (2.7%), and macrophage (φ; 0.6%).

(H–M) Expression of endocrine hormones (H) Ins1, (I) Ins2, (J) Sst, (K) Ppy, (L) Ghrl, and (M) Gcg across clusters.
population, there was little progression toward other cell fates, suggesting terminal differentiation (Figure 3B). Interestingly, a subset of acinar cells showed progression toward the trunk cell fate (Figure 3B). Within the trunk cluster, there were three root points determined using a Markov random-walk model (Figure 3C). From these roots, the cells transitioned through the EP state toward \(\alpha\) and \(\beta\) cells (Figure 3B). This was also confirmed by the endpoints of differentiation (Figure 3D). The same method was used to determine pseudotime of trunk, acinar, ductal, Endo, \(\alpha\), \(\beta\), and \(\delta\) cells at E18.5 (from Figure 1). For this analysis, a reduced dataset of 4,263 cells was visualized using t-SNE (Figure 3E). Unlike at E15.5, cells at E18.5 had less overall velocity, suggesting more terminally differentiated cells (Figure 3F). Within the Endo population there was a single differentiation root (Figure 3G) that bifurcates toward two cell types, the \(\alpha\) cells and a subpopulation of Endo cells. This is consistent with a population of EPs that serve as a root for differentiation and suggests that they either differentiate toward an endocrine cell lineage or remain as Endo cells. Together, these data highlight the heterogeneity of cell fate decisions during mouse pancreas development.

**Characterization of Endocrine Cell Population at E18.5**

To further investigate the heterogeneity of embryonic endocrine cells, the E18.5 green library was studied. Following filtering, 3,382 cells were visualized using a t-SNE plot and the following cell types were identified based on gene expression (Table S6): two \(\beta\) cell populations...
Figure 4. Characterization of Endocrine Cells in E18.5 Green Cells

(A) t-SNE plot of 10 cell clusters from E18.5 green cells: endocrine progenitors (EP; 7.0%), α cells (α; 18.2%), two β-cell clusters (β1, 23.4%, and β2, 11.8%), δ cells (δ; 16.7%), Ghrl-cells (ε; 8.0%), pancreatic polypeptide cells (PP; 6.9%), S-phase cells (S; 2.0%), mitotic cells (M; 2.4%), and exocrine cells (3.6%).

(B) Top ten differentially expressed genes in the β cell clusters: β1 (red), β2 (green), M (purple), and S (pink).

(C) Violin plots of average gene expression of Ins1, Ins2, Nkx6-1, Pdx1, Mafb, Scl2a2, G6pc2, Npy, and Gadd45a in β1 (red), β2 (green), M (purple), and S (pink).

(β1, 23.4%, and β2, 11.8%), α cells (18.2%), δ cells (16.7%), ε cells (8.0%), pancreatic polypeptide cells (PP; 6.9%), exocrine cells (3.6%), S-phase cells (2.4%), and mitotic cells (2.0%) (Figure 4A). The cell cycle phase of individual cells in the E18.5 green library was determined. As expected, the S-phase cluster contained both S- and G2/M-phase cells, while the M cluster had G2/M-phase cells (Figure SSA). To confirm the cell identities, the expression of endocrine hormones was determined in single cells. Both Ins1 and Ins2 were found in the cells of β1, β2, S, and M clusters, while Gcg expression was specific to the α cell population (Figure SSB). The δ cell cluster expressed Sst, the ε cells expressed Ghrl, and PP cells contained Ppy transcripts (Figures S5B and S5C).
To understand the heterogeneity within the β cell populations the expression of the top ten genes for β1 (pink), β2 (green), S-phase (pink), and M-phase cells (purple) was determined (Figure 4B). Most cells were in the β1 cluster; expressed genes involved in glucose metabolism, including Slc2a2 and G6pc2; and expressed high levels of Ins1 and Ins2 (Figure 4B and S5B). The top ten differentially expressed genes of the β2 cluster included the progenitor markers Nnat and Ppp1r1a, suggesting a less mature β cell population (Figure 4B). The other Ins-expressing cells are located within the S- and M-phase clusters. The S cluster expressed genes specific to the S phase, such as Cdk1, suggesting these cells represent a small (2.0%) population of β cells undergoing DNA replication (Figures 4B, S5A, and S5B). In the M cluster (2.4%), the cells expressed the histone genes Hist1h2bc and Hist1h1c, suggesting that these cells are undergoing mitosis (Figures 4B, S5A, and S5B).

Previous studies in adult β cells suggest proliferation is accompanied by a decrease in the function and maturation of β cells (Puri et al., 2018; Szabat et al., 2016). To understand if a similar process occurs during mouse β cell development, we profiled the expression of several β cell maturity and progenitor markers in the β1, β2, S-phase, and M-phase populations of cells (Figure 4C). The β2 cluster contained a subset of cells with lower Ins1, Ins2, G6pc2, and Npy, consistent with a less functional cell state (Figure 4C). In addition, this cluster showed an upregulation of genes expressed in immature progenitor cells: Nkx6-1, Pdx1, Mafb, and Gadd45α (Figure 4C). Interestingly, the cells of the S and M cluster exhibit a gene expression profile similar to that of the β1 cluster, suggesting that in the embryonic state proliferation does not reduce maturation (Figure 4C).

**Single-Cell Transcriptome of NEUROG3 Lineage during hESC Differentiation**

To profile the transcriptome of human EPs, a CyT49 NEUROG3-2A-eGFP hESC reporter line (N5-5) was used (Krentz et al., 2017). N5-S cells were differentiated using the protocol of Rezania et al. (Rezania et al., 2014) and collected for scRNA-seq before the transition to stage 6 (S6D1), when the number of GFP+ cells peaks (Figure S6A). The differentiation that was used for library generation had 23.9% GFP+ cells at S6D1 (Figure S6B). After filtering, 4,462 GFP+ cells were visualized using t-SNE, revealing nine clusters (Figure 5A). These clusters can be classified as five cell types based on gene expression: EPs (39.1%), polyclonal endocrine (Endo; 42.0%), duct (6.0%), liver (8.2%), and an unknown cell type (4.7%) (Figure 5A).

Further examination of the expression of endocrine-specific genes supported these cell classifications. Although all cells expressed GFP protein, only a few cells localized to the EP clusters expressed eGFP and NEUROG3 transcripts (Figure 5B). NEUROD1, a direct target of NEUROG3 (Huang et al., 2000), was widely expressed throughout the EP and endocrine clusters (Figure 5B). To understand how the three EP clusters differ, the top ten genes that are specific for each individual cluster were determined. The largest EP cluster (EP1) expressed NEUROG3, while EP2 contained genes that are associated with serotonin production, including TPH1 and FEV (Ohta et al., 2011) (Figure 5D). In EP3, the GAST gene was upregulated, consistent with previous reports of GAST induction in INS+ cells during hESC differentiation (Suissa et al., 2013) (Figure 5D).

The endocrine cell population was made up of hormone+ cells, many of which co-expressed multiple hormones, including GCG, SST, and INS (Figure 5C), consistent with a previous study characterizing hESC-derived endocrine cells using single-cell qPCR (Petersen et al., 2017). At the protein level, most cells were single-hormone positive, but a small proportion of double-hormone-positive cells was detected (Figure S6C). Of the three hormones, INS was the most abundantly expressed and can be detected in EP and endocrine cells (Figure 5C). Differential gene expression analysis between the three endocrine clusters revealed an enrichment of β cell genes ERO1β (Zito et al., 2010) and SLC30A8 (Davidson et al., 2014) in Endo1, suggesting these cells were differentiating β cells (Figure 5D). The Endo2 cluster appeared fated toward the α cell based on expression of GCG, PEMT (Segerstolpe et al., 2016), and IRX2, while the expression of SST and HHEX in Endo3 is suggestive of the δ cell fate (Figure 5D).

**Comparison of hESC-Derived Endocrine Cells to Mouse and Human Endocrine Cells**

To understand the “developmental age” of hESC-derived endocrine cells, the single-cell transcriptomes of S6D1 GFP+ cells were compared with mouse endocrine cells.
using scmap (Kiselev et al., 2018). This method allows comparison of multiple datasets by projecting a cell onto a reference dataset, inferring cellular identity based on what reference cell type it transcriptionally resembles. First, individual hESC-derived cells were mapped to E15.5 and E18.5 mouse endocrine cell types based on transcriptional similarity. Of the total number of hESC-derived EP cluster cells, 55.2% resembled mouse EPs and 14.9% resembled Endo cells (Figure 6A). Interestingly, 92.9% of all hESC-derived AFP-expressing cells and 93.7% of all hESC-derived duct cells mapped to mouse EP cells, suggesting that these populations represent an EP-like cell (Figure 6A). Of the endocrine cluster cells, 32.4% mapped with mouse embryonic ß cells, suggesting that most hESC-derived endocrine cells were fated to the ß cell lineage.

To understand how hESC-derived cells compare with adult human endocrine cells, we next compared hESC-derived cells with human islet cells (Baron et al., 2016). As expected, most cells of the three EP clusters were unassigned, as EPs are found only in embryonic pancreas (Figure 6B). However, the cells that do map within the EP clusters resemble ß cells, ð cells, and ductal cells (Figure 6B). Within the Endo clusters, 61.1%, 41.3%, and 29.3% of Endo1, Endo2, and Endo3 mapped to the ß cell lineage, respectively (Figure 6B). Consistent with the expression of GCG and SST (Figure 5C), 9.6% of Endo1 and 16.3% of Endo2 cells resemble ß cells, while 27.9% of Endo3 cells map to ð cells (Figure 6B). Interestingly, the three non-endocrine hESC-derived cell types, AFP, duct, and unknown, mapped to ductal cells in the adult dataset (Figure 6B). Taken together, 74.3% of assigned cells mapped to an endocrine lineage, with the vast majority mapping to the human ß cells, suggesting the current differentiation protocols generate endocrine cells that share transcriptomic similarity to human ß cells (Petersen et al., 2017).

**DISCUSSION**

scRNA-seq allows for the discovery of cell types and cell-state-specific genes, and the appreciation of cellular heterogeneity within a population. Here, scRNA-seq was used to generate a resource of single-cell transcriptomes from 6,905 E15.5 embryonic pancreatic cells, 6,626 E18.5 embryonic pancreatic cells, and 4,462 hESC-derived NEUROG3-2A-eGFP cells. Several unique observations were made, including genes that may regulate endocrine cell formation and previously unidentified populations of
cells generated in hESC differentiations. These data are publicly available online (https://lynnlab.shinyapps.io/embryonic_pancreas) and will serve as a single-cell gene expression resource for mouse embryonic pancreas and hESC-derived endocrine cells.

Comparison of E15.5 and E18.5 EP cells resulted in a list of potential markers, including two genes, Btg2 and Gadd45a, that are involved in neural development. Pro-neural proteins induce expression of Gadd45g (de la Calle-Mustienes et al., 2002; Huang et al., 2010), resulting in Gadd45-dependent cell cycle exit by upregulation of the cell cycle inhibitor Cdkn1a (Hildesheim and Fornace, 2002) and direct interaction with Cdk1/CyclinB (Zhan et al., 1999). While Gadd45 proteins are implicated in pancreatic cancer (Hildesheim and Fornace, 2002), their role in pancreas development has not been investigated. The expression of Gadd45a in Neurog3+ EP cells suggests that it may play a role, along with Cdkn1a, in regulating cell cycle exit.

Btg2, also known as Tis21, is a negative regulator of the cell cycle that inhibits transcription of CyclinD1, preventing the G1-S transition (Canzoniere et al., 2004; Guardavaccaro, 2000). Deletion of Btg2 in the adult dentate gyrus shortens G1 length in progenitor cells and prevents their terminal differentiation (Farioli-Vecchioli et al., 2009). This is thought to be caused in part by the direct binding of Btg2 to the Id3 promoter. Id proteins bind class A basic-helix-loop-helix (bHLH) proteins, which are obligate heterodimerization partners of class B bHLH TFs like Neurog3. By sequestering class A bHLH proteins and preventing their association with pro-neural bHLH TFs, Id acts to prevent terminal differentiation (Lyden et al., 1999). Btg2 may also act to inhibit Id3 transcription in the pancreas, allowing for the activation of pro-endocrine genes, including Neurog3.

The liver, like the pancreas, is derived from the foregut endoderm. The region of the endoderm that gives rise to the liver can also form the ventral pancreas (Tremblay and Zaret, 2005). One of the mechanisms that controls the decision between liver and pancreas is the secretion of fibroblast growth factors by the cardiac mesoderm, which permits the formation of liver while preventing ventral pancreas formation (Zaret et al., 2008). The similar developmental origins of the pancreas and the liver make the unintended generation of liver cells during hESC differentiations toward pancreas likely. However, finding liver cells downstream of NEUROG3 is surprising. It is possible that, like in the mouse, a small population of hESC-derived pancreatic endoderm cells have low transcription of NEUROG3 that is not sufficient to induce the endocrine lineage (Bechard et al., 2016). Alternatively, this may also be due to precocious activation of NEUROG3 during early stages of the differentiation before cells are competent to become endocrine cells (Russ et al., 2015). Using a NEUROG3 lineage-tracing hESC line, it would be interesting to investigate the plasticity of cells that activate NEUROG3 transcription. Our previous studies suggest that NEUROG3 protein in hESC differentiations is hyperphosphorylated (N.A.J.K. and F.C.L., unpublished data), resulting in rapid degradation. Efforts to stabilize NEUROG3 protein may prevent the unintended formation of other endodermal cell types, including liver cells.

In conclusion, the single-cell transcriptomes of mouse pancreatic progenitors, EPs, and endocrine cells at E15.5 and E18.5 as well as NEUROG3-expressing cells derived from hESCs were characterized. These data are a resource for developmental biologists interested in studying heterogeneity in the developing mouse pancreas and for stem cell researchers aiming to improve current differentiation protocols for generating β-like cells.

## EXPERIMENTAL PROCEDURES

### Animals

Mice were housed on a 12-hr light-dark cycle in a climate-controlled environment according to protocols approved by the University of British Columbia Animal Care Committee. Rosa26<sup>cmT/mG</sup> (Stock No. 007576) (Muzumdar et al., 2007) and Neurog3-Cre (Stock No. 005667) mice were purchased from The Jackson Laboratory.

### Preparing Cells for Single-Cell RNA Sequencing

For mouse studies, Neurog3-Cre; Rosa26<sup>cmT/mG</sup> embryos were collected on E15.5 and E18.5 and dissected on ice. To generate single cells, embryonic pancreases were incubated in 2 mL of pre-warmed 37°C 0.25% trypsin with mild agitation for 8 or 20 min for E15.5 and E18.5 pancreases, respectively. To stop digestion, 1 mL of cold fetal bovine serum (FBS) and 2 mL of cold PBS were added and mixed by inversion, followed by filtering through a 40-µm nylon filter. Cells were then centrifuged at 4°C for 5 min at 200 x g. After the supernatant was aspirated, the cells were resuspended in cold 2% FBS in PBS, placed on ice, and immediately sorted by a 12-well plate. Following 5 min at 37°C, 500 µL of 2% BSA MCDB medium was added to each well and the cells were transferred to a 15-mL conical tube. Cells were centrifuged for 5 min at 200 x g, washed once with PBS, and resuspended in 350 µL of ice-cold PBS. GFP+ cells were sorted into stage 5 medium with 500 µL of Y-27632 dihydrochloride using a Beckman Coulter MoFlo Astrios. scRNA-seq libraries were generated with 10× Genomics Chromium pipeline (see Supplemental Experimental Procedures for details).
Data Analyses
Following sequencing, data were analyzed using publicly available software programs and R pipelines. First, cellranger mkfastq (10x Genomics) was used to generate FASTQ files from the raw sequencing data. Next, cellranger count aligned FASTQ files to reference mouse (GRCm38) and human genomes (GRCh38) and generated single-cell gene counts. To determine counts for the lineage-tracing transgenes, the mouse and human reference genomes were annotated to include the tdTomato and/or eGFP sequences. Cellranger aggr was used to combine data from multiple libraries to generate E15.5 aggr (E15.5 red and E15.5 green and yellow), E18.5 aggr (E18.5 red, E18.5 yellow, and E18.5 green), and E18.5 endocrine (E18.5 yellow and E18.5 green), using the normalization parameter set to “mapped” to normalize sequencing depth across libraries.

Two additional R pipelines were used to filter out cells that did not meet the quality control standard. Scater (https://bioconductor.org/packages/release/bioc/html/scater.html) was used to discard cells: (1) with fewer than 1,000 transcripts, (2) with counts (transcripts/cell) greater than 3 median absolute deviation (MAD) away from the median (removing potential doublets and debris), (3) with genes (genes/cell) greater than 3 MAD away from the median (to remove low-abundance genes or genes with high dropout rate), and (4) with mitochondrial DNA transcripts/total transcripts greater than 3 MAD away from the median (McCarthy et al., 2017). The runPCA automatic filtering algorithm was used to filter out genes and cells based on principal-component analysis (PCA). Finally, a gene must be expressed in at least three cells to be included in the downstream analysis.

This quality control dataset was then analyzed using the Seurat v.2.0 pipeline (http://satijalab.org/seurat/), another R toolkit for single-cell transcriptomics (Butler et al., 2018). Seurat was used to remove common sources of variation, including number of genes (each cell must express a minimum of 500 genes), number of counts (each gene must be expressed in a minimum of three cells), and cell cycle phase. The Seurat object was log normalized with a default size factor of 10,000, and data were scaled while regressing out cell-cycle-dependent changes in gene expression. To determine variable genes, mean expression was used with the LogVMR dispersion function and imputed into PCA. The first 15 principal components were used to cluster based on a shared nearest-neighbor modularity optimization-based clustering algorithm with default parameters and resolution of 0.6. t-SNE embedding using the first 15 principal components was used to visualize data in reduced dimensions. To identify differentially expressed genes in each cluster, the Seurat function FindAllMarkers was used. For a gene to be differentially expressed in a cluster it must be expressed by at least 10% of cells, have a log-fold change greater than 0.25, and reach statistical significance of an adjusted p < 0.05 as determined by the Wilcox test.

Human-Mouse Comparison
scmap was used to analyze the similarity between hESC-derived cells and mouse embryonic pancreas or human islet cells (Kiselev et al., 2018). For the comparison with embryonic mouse cells, endocrine cells from the E15.5 yellow and green and E18.5 yellow and green libraries were used. For the comparison with human islets, four human islet libraries were used (Baron et al., 2016). Scater was used to normalize gene expression by library size and perform a log2 transformation of the data, and libraries were subset to include only the expression of shared genes across the datasets. The top 500 genes for each dataset on the linear model of log(expression) versus log(dropout rate) were selected using selectFeatures. The mean expression, or centroid, of the selected genes was calculated for each cluster in the reference datasets using indexCluster with default parameters. The comparison between a query dataset (hESC-derived) to each reference library was carried out by scmapCluster with a minimum similarity threshold of 0.5.

Pseudotime Analysis
Pseudotime analysis was performed on the post-scater and Seurat dataset using RNA Velocity (Manno et al., 2018). First, the expression of spliced and unspliced transcripts of each gene was determined. Cells in the bottom 0.5% of the total unspliced transcript count were removed. Next, the following filtering was performed to discard genes: (1) with fewer than 40 total spliced counts across all cells, (2) with fewer than 30 cells expressing the spliced variant, (3) with fewer than 25 total unspliced counts across all cells, (4) with fewer than 20 cells expressing the unspliced variant, or (5) with an average spliced variant expression of less than 0.08 or average unspliced variant expression of 0.01 in at least one cluster. Genes were then ranked using the coefficient of variation versus mean metric, selecting the top 3,000 genes as features. The velocity estimation was performed in Python using the gene relative estimation method. Gamma fit was performed using cell nearest-neighbor pooling with k = 500. The resulting velocity estimates were projected onto the t-SNE embedding obtained in Seurat.

ACCESSION NUMBERS
Sequencing data are available under GEO Accession GSE120522.

SUPPLEMENTAL INFORMATION
Supplemental Information includes Supplemental Experimental Procedures, six figures, seven tables, and one data file and can be found with this article online at https://doi.org/10.1016/j.stemcr.2018.11.008.

AUTHOR CONTRIBUTIONS
Conceptualization, N.A.J.K., E.E.X., S.S., and F.C.L.; Methodology and Investigation, N.A.J.K., M.L., E.E.X., S.L.J.S., A.M., S.S., and F.C.L.; Writing, N.A.J.K. and F.C.L.; Funding Acquisition, F.C.L.

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