Cell Metabolism

Heterogenous impairment of α cell function in type 2 diabetes is linked to cell maturation state

Graphical abstract

Highlights

- Glucose suppresses α cell exocytosis by inhibiting P/Q-type Ca²⁺ currents
- Patch-seq links maturation, respiration, and receptor expression to α cell function
- Dysfunction of α cells associates with a “β cell-like” electrophysiological signature
- Impaired exocytosis occurs in α cells enriched for lineage and immaturity markers

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In brief

In diabetes, glucagon secretion from pancreatic α cells is dysregulated. Dai et al. examined electrical and transcriptomic α cell phenotypes and found that dysfunction in type 2 diabetes is linked to cell maturation state and impaired α cell identity. Notably, a subset of α cells enriched for lineage markers appears uniquely susceptible to dysfunction.
Heterogenous impairment of α cell function in type 2 diabetes is linked to cell maturation state


SUMMARY

In diabetes, glucagon secretion from pancreatic α cells is dysregulated. The underlying mechanisms, and whether dysfunction occurs uniformly among cells, remain unclear. We examined α cells from human donors and mice using electrophysiological, transcriptomic, and computational approaches. Rising glucose suppresses α cell exocytosis by reducing P/Q-type Ca²⁺ channel activity, and this is disrupted in type 2 diabetes (T2D). Upon high-fat feeding of mice, α cells shift toward a “β cell-like” electrophysiological profile in concert with indications of impaired identity. In human α cells we identified links between cell membrane properties and cell surface signaling receptors, mitochondrial respiratory chain complex assembly, and cell maturation. Cell-type classification using machine learning of electrophysiology data demonstrated a heterogenous loss of “electrophysiologic identity” in α cells from donors with type 2 diabetes. Indeed, a subset of α cells with impaired exocytosis is defined by an enrichment in progenitor and lineage markers and upregulation of an immature transcriptomic phenotype, suggesting important links between α cell maturation state and dysfunction.

INTRODUCTION

In concert with reduced insulin secretion from pancreatic β cells in type 2 diabetes (T2D), disrupted glucagon secretion from α cells contributes to hyperglycemia and impaired hypoglycemia counter-regulation (Girard, 2017). While insulin and glucagon secretion are both dependent on electrical excitability and Ca²⁺-dependent exocytosis, the nature of the ion channels involved and their roles and the impact of glucose-stimulation differ in α and β cells. Insulin granule exocytosis is linked to Ca²⁺ entry via L-type Ca²⁺ channels, whereas glucagon secretion is coupled to P/Q-type Ca²⁺ channels (De Marinis et al., 2010). Also, Na⁺ channels play a more prominent role in glucagon secretion (Barg et al., 2000; Göpel et al., 2000; Ramracheya et al., 2010) and differ in their regulation between these cell types (Zhang et al., 2014). In rodents, such differences can distinguish α from β cells, either by Na⁺ current properties (Zhang et al., 2014) or “electrophysiological fingerprints” (Briant et al., 2017).

Similar to β cells, the excitatory and secretory machinery in α cells, including ion channel activities (Huang et al., 2011a),
Ca\textsuperscript{2+} responses (Le Marchand and Piston, 2010; Reissaus and Piston, 2017; Shuai et al., 2016), glucagon content (Zadeh et al., 2020), and exocytotic capacity (Huang et al., 2011b), is heterogeneous. Indeed, the intracellular Ca\textsuperscript{2+} response of \( \alpha \) cells varies, with some suppressed by glucose and others activated (Shuai et al., 2016). This suggests a heterogeneity among these cell types, which is supported by single-cell transcriptomics (Camunas-Soler et al., 2020; Korsunsky et al., 2019). Recent reports suggest a subpopulation of \( \alpha \) cells (or "\( \alpha \)-like cells") that are pro-liferative. In mice these are identified by Slc38a5, which encodes an amino acid transporter (Kim et al., 2017); in humans these may be identified by the presence of ARX and cytosolic Sox9 (Lam et al., 2018). These cells could account for \( \alpha \) cell hyperplasia upon glucagon-receptor antagonism and may be a source of new \( \beta \) cells (van der Meulen et al., 2017).

While little evidence so far suggests that \( \alpha \) cells dedifferentiate in diabetes, they can trans-differentiate in rodents following severe \( \beta \) cell loss (Thorel et al., 2010) or genetic manipulation of transcription factor expression (Chakravarthy et al., 2017; Matsuoka et al., 2017). Interestingly, \( \alpha \) cells may show more plasticity than \( \beta \) cells (Bramswig et al., 2013), as they appear to exist in distinct states characterized by chromatin accessibility at promoters for GCG, functional genes such as ABCB8, and at sites enriched in motifs for transcription factors of the RFX, GATA, and NEUROD families, among others (Chiou et al., 2021). In type 1 diabetes, the expression of \( \alpha \) cell markers is reduced (Brissova et al., 2018) and in T2D \( \alpha \) cells express an immature transcriptomic profile (Avrahami et al., 2020).

We hypothesized that this plasticity influences \( \alpha \) cell membrane function, contributing to dysfunction in T2D. We used correlated electrophysiological and single-cell RNA seq (patch-seq) of \( \alpha \) cells from human donors and mice to define a cell-autonomous glucose regulation of \( \alpha \) cell Ca\textsuperscript{2+} channel activity and exocytosis that is associated with \( \alpha \) cell maturation state and identified putative regulators of glucagon secretion. These include the mitochondrial respiratory chain complex and numerous cell surface receptors. In mice, high-fat feeding prompts some \( \alpha \) cells to adopt a "\( \beta \) cell-like" electrical profile and impaired identity. Similarly, in human T2D, \( \alpha \) cells enriched for markers of mitochondrial function and endocrine lineage such as NEUROD1, ISL1, NKX2-2, and ARX exhibit a selective induction of an immature transcriptional profile, impaired electrophysiological phenotype, and dysregulated exocytosis. This suggests an important link between \( \alpha \) cell maturation, identity, and dysfunction in T2D.

**RESULTS**

**Glucose-mediated suppression of \( \alpha \) cell exocytosis is disrupted in T2D**

In \( \beta \) cells, glucose metabolism amplifies Ca\textsuperscript{2+}-triggered exocytosis and insulin secretion (Ferdaoussi et al., 2015; Gembl et al., 1992; Sato et al., 1992), which is linked to the activation of L-type Ca\textsuperscript{2+} channels (Barg et al., 2001; Bokvist et al., 1995; Wiser et al., 1999). We examined the impact of glucose on glucagon exocytosis in \( \alpha \) cells from donors with no diabetes (ND) or with T2D (Table S1) following dispersion to single cells and identification by glucagon immunostaining (Figure 1A). In ND \( \alpha \) cells, exocytosis was highest at low glucose (1 mM) and was suppressed by increasing glucose (to 5–20 mM) (Figures 1B and S1). In \( \alpha \) cells from donors with T2D, exocytosis was ~75% lower at 1 mM glucose than in \( \alpha \) cells from ND donors, and elevating glucose levels exerted a modest increase (Figures 1B and S1A). The differences between ND and T2D cells, also seen when grouped by donor (Figure S1B), are consistent with a recent live-cell imaging study (Omar-Hmeadi et al., 2020). In ND \( \alpha \) cells, voltage-dependent Ca\textsuperscript{2+} channel activity was lower at elevated glucose than in \( \alpha \) cells from donors with T2D, here recorded using Ba\textsuperscript{2+} as a charge carrier (Figures 1C and S1C), and the relationship between Ca\textsuperscript{2+} entry and exocytosis was not altered (Figure S1D).

The majority of the human \( \alpha \) cell Ca\textsuperscript{2+} current is mediated by L-type and P/Q-type channels (Ramacheya et al., 2010), and the latter are directly linked to glucagon exocytosis (Dai et al., 2014; Ramacheya et al., 2010, 2018). The low \( \alpha \) cell exocytosis at elevated glucose could be reversed by the P/Q-type Ca\textsuperscript{2+} channel activator GV-58 (Tarr et al., 2012), which delays Ca\textsuperscript{2+} current inactivation (Figure 1D). The P/Q-type channel blocker agatoxin inhibited exocytosis from ND \( \alpha \) cells at 1 mM glucose while the L-type channel blocker isradipine did not (Figure 1E). Finally, the greater P/Q-type Ca\textsuperscript{2+} channel activity seen in ND \( \alpha \) cells at 1 mM glucose was largely absent in T2D (Figures 1F and S1A) but still contributed to the modest glucose-dependent increase in exocytosis (Figure 1G). Thus, in ND \( \alpha \) cells increasing glucose inhibits P/Q-type Ca\textsuperscript{2+} channels to limit exocytosis, an effect that requires intact mitochondrial function (Figures S1E and S1F). In T2D \( \alpha \) cells, Ca\textsuperscript{2+} channel activity at low glucose is reduced and increasing glucose facilitates exocytosis (Figure 1H).

Patch-seq highlights a role for the mitochondrial respiratory chain in \( \alpha \) cell exocytosis and suggests poor responsiveness in immature \( \alpha \) cells

We performed patch-seq in \( \alpha \) cells, some initially pre-incubated at low glucose for 1 h. However, exocytosis was low after an extended time at 1 mM glucose due to depletion of glucagon granules (Figure S2). These pre-incubated cells were excluded from further analysis, and we performed transcriptome-wide analyses with exocytosis in 400 human \( \alpha \) cells (from 31 donors) exposed acutely to 1 and 10 mM glucose (Figure 2A). ND \( \alpha \) cells at 1 mM were enriched for positively correlated genes while negatively correlated genes were enriched at 10 mM glucose (Figure 2B; Table S2). Gene set enrichment analysis (GSEA) using Z scores of transcripts found in >20% of cells highlighted pathways associated with elevated \( \alpha \) cell exocytosis at 1 mM glucose and lower exocytosis at 10 mM glucose (Figure 2C). Consistent with a role for metabolism in determining \( \alpha \) cell responsiveness, we found mitochondrial respiratory chain complex assembly as a positive correlate to exocytosis at low glucose and a negative correlate at high glucose (Figures 2C and 2D). A separate over-representation analysis (ORA) yielded similar results, highlighting the electron transport chain in the glucose regulation of \( \alpha \) cell exocytosis (Figure S3A).

Ca\textsuperscript{2+} channel transcripts were not reduced in \( \alpha \) cells of T2D donors compared with those from ND donors (Figures S3B and S3C). Hyperglycemia may induce \( \alpha \) cell mitochondrial dysfunction (Knudsen et al., 2019), and we saw a modest but
significantly higher mitochondrial respiratory chain complex transcript expression in \( \beta \) cells of T2D donors compared with \( \beta \) cells from ND donors (Figures S3D and S3E). Intriguingly, transcripts associated with endocrine development appeared as anti-correlates of exocytosis at 1 mM glucose (Figures 2C, 2E, and 2F), suggesting a role for cell differentiation state in the responsiveness of \( \beta \) cells. In \( \beta \) cells from donors with T2D, the effect of glucose on the distribution of transcriptome-wide correlations was reversed, such that high glucose associated with more positively correlated genes (Figure 2G; Table S3). GSEA of these correlations highlighted a role for cell development state in the inappropriately high \( \beta \) cell exocytosis at 10 mM glucose in T2D (Figure 2H). Several leading-edge transcripts enriched in these pathways, including transcription factors important for pancreatic endocrine maturity like \( \text{GATA6} \), \( \text{PAX6} \), \( \text{RFX6} \), and \( \text{RFX3} \), overlap with those that correlated with inappropriately low exocytosis in \( \beta \) cells of ND donors at 1 mM glucose (Figures 2F and 2I).

Glucose control of mouse \( \beta \) cell exocytosis and impaired “electrophysiological identity” following high-fat feeding

Following an \(~10\)-min exposure to 1 mM glucose, depolarization of mouse islets with 20 mM KCl elicited a transient stimulation of glucagon release that was blunted in islets kept at 5 mM glucose (Figure 3A), consistent with the glucose-dependent suppression of \( \beta \) cell exocytosis. These differences could not be explained by the indirect paracrine effects of insulin, as insulin secretion evoked by KCl was slightly higher at 1 mM glucose than at 5 mM glucose (Figure 3B). Similar to human \( \beta \) cells, increasing glucose suppressed exocytosis in mouse \( \beta \) cells (Figure 3C). Although the physiological impact of glucose on glucagon secretion involves key paracrine signals (Briant et al., 2016), the suppression of exocytosis by increasing glucose was similar when cells were seeded at 10% of normal density (Figure 3D) and required glucose metabolism, as the non-metabolizable analog 2-deoxyglucose (2-DG) did not
mimic the effect of glucose (Figure 3E). The glucose suppression of Ca\textsuperscript{2+} entry and exocytosis could be prevented by rotenone, but not antimycin A (Figure 3F), suggesting a signal at or between mitochondrial respiratory chain complexes I and III. This could include reactive oxygen species/H\textsubscript{2}O\textsubscript{2} produced by complex I via reverse electron transport (Onukwufor et al., 2019) or by complex III toward the cytosol (Muller et al., 2004). Indeed, direct intracellular dialysis of H\textsubscript{2}O\textsubscript{2} suppressed...
α cell Ca\(^{2+}\) influx and exocytosis in mouse (Figure 3F) and human (Figures S1E and S1F) α cells.

Low-glucose stimulation of glucagon secretion was enhanced in mice fed a high-fat diet (HFD) for 12–14 weeks compared with age-matched controls (18–20 weeks; Figure 4A), similar to what we have shown previously (Kellard et al., 2020). There was no difference in insulin secretion at 5 and 1 mM glucose but the response to high K\(^+\) was reduced in mice fed an HFD compared with controls (Figures S1E and S1F).

Physiological fingerprint modeling links human α cell behavior and cell phenotype

Next, we compiled data from islet cells of 67 donors collected at 1, 5, and 10 mM glucose, identified either by scRNA-seq or immunostaining (Figure 5A). Unlike in mice, but similar to previous reports (Braun et al., 2008; Ramracheya et al., 2010), Na\(^+\) channel properties are similar between human α and β cells and could not be reliably used alone to distinguish these cell types (Figures 5B, 5C, and S4C). In ND α cells, the peak Na\(^+\) current significantly correlated with genes known to impact α cell activity (Figure 5D). Top positive correlates include genes involved in α cell or islet function, including Na\(^+\) and Ca\(^{2+}\) channels (SCN3A, SCN3B, and CACNA1A), α and islet cell lineage transcription factors.
and random forests (Camunas-Soler et al., 2020) models that types,” similar to the logistic regression (Briant et al., 2017) we had used previously. In those models, however, cell size is known to regulate a GC abundant but included the vitamin D-binding protein (Figure 6 B) to characterize human channel activity (Figures 5 E and S5C). We validated the ability of agonists of some of these to increase human cells makes it impossible to use these alone to interrogate shifts in human α cell phenotype. We therefore developed “electrophysiological fingerprinting” classifier models (Figure 6A) that integrate Na+ current, Ca2+ current, and exocytosis measures (Figure 6B) to characterize human α cell “functional phenotypes,” similar to the logistic regression (Briant et al., 2017) and random forests (Camunas-Soler et al., 2020) models that we had used previously. In those models, however, cell size was the major predictor of α cell identity. Here, we generated three independent models for cell type classification: optimal ensemble classifiers that include (Model 1) or exclude (Model 2) cell size as an independent variable, and an extreme gradient boosting model that also excludes cell size, but with additional restrictions to donor age, donor body mass index (BMI), and organ cold ischemic time (CIT) applied to the training data (Model 3). These models distinguished ND α and β cells well (Figure 6A) and, unlike some underlying parameters, were unaffected by potential confounders, such as donor sex, BMI, CIT, cell culture time, and glucose concentration (Figure 6B).

Model output between 0 and 1 can be considered the probability of an electrophysiological profile matches that of an α cell, and we called this ρprobability (with $\rho_{\text{probability}} = 1 - \rho_{\text{probability}}$). Assignment of an ρprobability to all ND α and β cells for which we had patch-seq data, without an a priori knowledge of cell type, showed the effective separation of cell types and correlation with canonical markers (Figure 6C). All three models showed a significantly reduced $\rho_{\text{probability}}$ and ρprobability in β and α cells, respectively, from donors with T2D compared with cells from ND donors (Figure 6D). The expression of numerous transcripts correlated with a “loss of electrophysiological identity” in T2D α cells within one or more of the models, including lineage markers such as ISL1, NEUROD1, FEV, and RFX6 (Figure 6E; Table S7). GSEA using $\rho_{\text{probability}}$ correlation slopes as weighting for transcripts expressed...
in >20% of cells highlighted GO biological process terms related to mitochondrial respiratory chain complex assembly in all three models (Figure 6F), underscoring a link between increased respiratory chain complex expression and cell dysfunction.

Impaired functional identity and exocytosis in T2D α cells enriched in lineage and immaturity markers

Human α cells exist in states of variable chromatin accessibility at the GCG promoter and sites enriched with transcription factor motifs characterizing endocrine lineage and development (Chiou et al., 2021). Markers of α cell lineage are heterogenous within our dataset, with evidence for increased expression in T2D (Figures 7A, S6A, and S6B). Intriguingly, this also included ARX, which we confirmed at the protein level, along with MAFB and glucagon itself (Figure 7B). ARX-enriched (ARXhi) cells expressed consistently higher levels of progenitor transcripts like NEUROD1, FEV, GATA6, ISL1, NKX2-2, PAX6, and MAFC (Figures 7C and S6). Accordingly, ARX and these other markers did not correlate with exocytosis in α cells of ND donors but did correlate significantly with impaired exocytosis in T2D (Figure 7D). In separate scRNA-seq data (Avrahami et al., 2020) we confirmed the upregulation of a juvenile α cell profile (Arda et al., 2016), tissue development genes, and a β cell-like profile selectively in ARXhi α cells in T2D compared with α cells from ND donors (Figures 7E and S7). Finally, while the α cells with low ISL1, NKX2-2, NEUROD1, and ARX from donors with T2D had normal exocytosis at 1 mM glucose, α cells enriched in these transcripts showed impaired exocytotic function in T2D (Figure 7F).

DISCUSSION

Glucagon secretion is under the control of metabolic, paracrine, hormonal, and neuronal signals (El et al., 2020). Much
Figure 6. Electrophysiological fingerprints define a loss of “functional identity” in T2D α cells
(A) Classifier models trained on islet cell electrical properties of α and β cells from donors with no diabetes (ND) using optimizable ensemble or extreme gradient boosting (XGBoost) approaches identify cell types with high accuracy regardless of the inclusion or exclusion of cell size from training data; 80% of data was used for training; 20% of data was reserved for validation and generation of confusion matrices. tSNE plots show cell types determined by immunostaining or sequencing (left) and assigned probability scores (right).
(B) Ordinary least-squares multiple regression of electrophysiological properties of α cells from ND donors, model scoring, and donor/isolation variables.
(C) Calculated α probability (and β probability) values from Model 3 applied to cells collected for patch-seq, without a priori knowledge of cell type and correlation with canonical β cell (light blue) and α cell (pink) markers.
(D) β probability and α probability values derived from all three models, of all β and α cells from ND or T2D donors.
(E) Volcano plot of transcript correlations with Model 3 α probability values (slope/standard deviation) in α cells from donors with T2D. A negative correlation indicates transcripts associated with reduced α probability.
(F) Significantly enriched GO biological pathways from gene set enrichment analysis (GSEA) performed using α probability - transcript correlation slopes of α cells from donors with T2D in all three Models as weighting.

***p < 0.001 and ****p < 0.0001 as indicated within models using non-parametric Kruskal-Wallis test (D) followed by Dunn’s post-test to correct for multiple comparisons. Pink/red points in (E) indicate significance at p < 0.05. FDRs for pathways identified by GSEA in (F) were <0.05 unless indicated otherwise.
debate has centered on the question of whether glucose suppression of glucagon secretion is mediated via intrinsic, paracrine, or autonomic mechanisms (Gylfe, 2016), although it seems likely that α cells (like β cells) adeptly integrate multiple signals for precise physiologic control of glucagon. A role for direct glucose sensing is supported by the ability of glucose to modulate α cell ATP-sensitive K⁺ channels (Zhang et al., 2013), the impact of α cell glucokinase manipulation on in vitro cell function, and in vivo glucagon secretion and glucose homeostasis (Moede et al., 2020; Bahl et al., 2021; Basco et al., 2018). Here, we show that glucose suppresses human and mouse α cell exocytosis, consistent with a recent report where exocytosis was measured by live-cell imaging (Omar-Hmeadi et al., 2020). A role for α cell metabolism is supported by the effect of the non-metabolizable glucose analog 2-DG and the correlation of responses with mitochondrial respiratory complex assembly transcripts, particularly those of complex 1. Indeed, the ability of rotenone to block glucose suppression of α cell Ca²⁺ currents and exocytosis suggests a signaling role for reactive oxygen species and H₂O₂ generated by complex I. Consistent with this, intracellular application of H₂O₂ mimics the effect of glucose while reduced glutathione blocks the effect of glucose, and external H₂O₂ inhibits low-glucose-stimulated glucagon secretion (Allister et al., 2013).

The study by Omar-Hmeadi et al. (2020) showed a U-shaped response of isolated α cells to glucose, suggesting that glucose may increase α cell exocytosis under some conditions. We found some hints of this in our own data, including in α cells of donors with T2D. In ND α cells this may depend on the culture conditions. Following prolonged low-glucose preincubation α cells were larger, possibly due to continued glucagon granule fusion. Subsequent exocytotic responses were smaller when the α cells were maintained at 1 mM glucose, while acute glucose-stimulation then increased exocytosis. This is reminiscent of the glucose-dependent increase in glucagon secretion from purified α cells (Olsen et al., 2005). Thus, glucose acutely suppresses P/Q-type Ca²⁺ channels via a complex-I-dependent mechanism to limit depolarization-induced α cell exocytosis, while at the same time facilitating glucagon granule priming or docking to maintain the availability of granules. Altogether, these processes will “tune” α cell responsiveness to a myriad of paracrine, endocrine, and neuronal inputs.

Figure 7. A role for maturation state in α cell dysfunction in T2D
(A) Heterogeneous expression of transcript makers for islet cell lineage and α cell maturity in 980 patch-seq α cells. (B) ARX and MAFB protein expression in α cells at the protein level in situ by immunostaining. Violin plots show the relative levels of GCG, ARX, and MAFB expressed in ARX⁺ and ARX⁻ α cells. (C) Probability values from the three separate classifier models in ND and T2D α cells separated by high and low expression of ARX (see also Figure S6). (D) Correlation of exocytosis in ND and T2D α cells with α cell lineage and identity markers. Bars to the left of the centerline indicate correlation with low exocytosis at the given glucose concentration. (E) In a separate human islet single-cell dataset (Avrahami et al., 2020), expression of a juvenile α cell gene set in ND and T2D α cells separated by low and high ARX expression. The heatmap displays relative expression levels as median log₂ FPKM values. (F) Total exocytosis in ND and T2D α cells at 1 mM glucose separated by low and high expression of ISL1, NEUROD1, NKX2-2, and ARX. *p < 0.05; **p < 0.01; ***p < 0.001; ****p < 0.0001 by two-way ANOVA followed by two-stage step-up method for estimation of FDR (C, D, and F) or Tukey post-test (B).
The P/Q-type $\text{Ca}^{2+}$ current seen at low glucose in $\alpha$ cells of ND donors appears absent in T2D but channel expression is not reduced; therefore, channel regulation appears disrupted. It seems possible that altered mitochondrial function in $\alpha$ cells in T2D could drive dysregulation of P/Q-type $\text{Ca}^{2+}$ channel activity, and indeed, mitochondrial dysfunction has been demonstrated in $\alpha$ cells from hyperglycemic mice (Knudsen et al., 2019) and mitochondrial morphology is altered in $\alpha$ cells from Western-diet-fed mice (Grubelnik et al., 2020). Genes involved in glucose metabolism appear downregulated in $\alpha$ cells from donors with type 1 diabetes (Brissova et al., 2018), although these may exhibit a more extreme phenotype than in T2D, with downregulation of several channels, exocytotic transcripts, and $\alpha$ cell identity markers. We do find an upregulation, particularly in ARX$^\text{hi}$ T2D $\alpha$ cells, of mitochondrial respiratory chain complex assembly transcripts and a clear correlation of this with the impaired $\alpha$ cell phenotype by fingerprint modeling; however, a demonstration of altered mitochondrial respiratory function and/or reactive oxygen species generation in $\alpha$ cells from donors with T2D remains to be shown.

The coupling of glucagon exocytosis to P/Q-type $\text{Ca}^{2+}$ channels, which we confirmed at low glucose, has been demonstrated previously (De Marinis et al., 2010; Ramracheya et al., 2010) and is similar to the direct coupling of insulin granule exocytosis to the activation of L-type $\text{Ca}^{2+}$ channels (Wiser et al., 1999). These and other electrophysiological properties are used to distinguish $\alpha$ and $\beta$ cells in rodents. Most commonly, mouse islet cell types are identified by a combination of cell size ($\alpha$ cells are smaller) and distinct properties of Na$^+$ current inactivation. In a model of genetically induced $\alpha$-to-$\beta$ trans-differentiation in mice, we reported a clear shift in electrophysiological phenotype consistent with the attainment of $\beta$ cell properties (Chakravarthy et al., 2017). Intriguingly, following high-fat feeding, mouse $\alpha$ cells undergo a negative shift in Na$^+$ current inactivation and convert from P/Q-to L-type $\text{Ca}^{2+}$ channel dependence of exocytosis. While we provide some evidence for impaired $\alpha$ cell identity, this is not associated with a clear transdifferentiation, as $\beta$ cell markers are not increased and cells maintain positive immunostaining for glucagon. The shift in Na$^+$ channel inactivation toward a “$\beta$ cell-like” phenotype, which could be related to changes in membrane composition (Godzagar et al., 2018), correlates with the expression of important $\alpha$ cell lineage and identity transcription factors, suggesting that the change in electrical phenotype occurs more readily in cells with higher levels of these markers.

A subset of human $\alpha$ cells, even from donors without diabetes, exist in an immature state and may suffer a further loss of mature identity in T2D (Avrahami et al., 2020), perhaps related to their greater epigenetic plasticity (Bramswig et al., 2013) and distinct states of chromatin accessibility (Chiu et al., 2021). We find that $\alpha$ cells from ND donors with inappropriately low exocytosis are enriched in transcripts and pathways associated with endocrine development (FOXO1, PAX6, RFX6, and others). In T2D we see no obvious loss of $\alpha$ cell transcription factors or upregulation of $\beta$ cell-defining transcription factors indicative of trans-differentiation per se. We do, however, find heterogeneity in many lineage markers, as reported previously by us (Camunas-Soler et al., 2020; Drigo et al., 2019) and by others (Li et al., 2016). In our dataset, most of these show a small but significant increase in T2D. To assess a shift in electrophysiological phenotype of these human $\alpha$ cells we could not use Na$^+$ current inactivation alone, as this feature overlaps with measurements from human $\beta$ cells. We therefore modified machine-learning approaches that we used previously to improve the identification of mouse (Briant et al., 2017) and human (Camunas-Soler et al., 2020) islet cells. We used exocytosis, Na$^+$ current, and Ca$^{2+}$ properties as training data for three separate models with similar results. Two of these excluded cell size to solely identify shifts in membrane “activity,” and all models accurately assigned probability values for $\alpha$ and $\beta$ cells irrespective of ambient glucose, culture times, and other important donor- and isolation-related parameters. Correlation of $\text{probability}$ values with transcriptomic data therefore emphasized pathways linked to a “loss of functional phenotype” in T2D and suggests that $\alpha$ cells with the most altered electrical phenotypes in T2D are those with higher expression of pancreatic endocrine lineage markers.

In this study, we define an impaired $\alpha$ cell phenotype in T2D that is largely restricted to a subgroup of cells expressing higher levels of markers that define not only $\alpha$ cell maturity but pancreatic endocrine lineage. At first look we would have expected that high levels of ARX should be indicative of mature $\alpha$ cell function but this is misleading given a strong overlap of ARX$^\text{hi}$ cells with transcripts more traditionally associated with an endocrine progenitor state such as NEUROD1, ISL1, FEV, and others. These cells indeed appear more “plastic” in their electrophysiological responses, which may predispose them to an altered electrical phenotype in T2D. We confirm in a separate dataset the “de-repression” of immature gene sets associated with a “juvenile” $\alpha$ cell phenotype and with tissue development restricted to ARX$^\text{hi}$ $\alpha$ cells in T2D. Thus, we demonstrate that all $\alpha$ cells are not equally impacted by disease and that a subset of $\alpha$ cells defined by their maturation state may be key drivers of impaired glucagon responses in T2D.

Limitations of study

While tempting to link impaired exocytosis in $\alpha$ cells from donors with T2D to an impaired responsiveness to hypoglycemia in vivo, this must be considered in the context of in situ $\alpha$ cell function, which will be impacted by the local environment and by paracrine or hormonal signals. Here, we studied single isolated $\alpha$ cells, and although we demonstrate a similar heterogeneity in ARX/MAFB protein expression in situ, $\alpha$ cell function may be different within the intact pancreas and would likely be impacted by architectural changes in the disease state. Encouragingly though, we find clear differences between ND and T2D $\alpha$ cells that persist in vitro, and among the novel findings of this study we also find well-established regulators of $\alpha$ cell function (such as GIPR). Additionally, both $\alpha$ cell function and transcript expression are likely dynamic and impacted by metabolic status or culture conditions. Indeed, preculture at low glucose can alter $\alpha$ cell exocytotic function. Our approach to “electrophysiological fingerprinting” addresses this in part since it is unaffected by glucose, time in single-cell culture (up to 3 days), or various donor-related parameters. Nonetheless, the relevance of possible dynamic shifts in $\alpha$ cell phenotype in T2D remains unclear. We do not know, for example, if exocytotic function would be restored if T2D $\alpha$ cells transition from an ARX$^\text{hi}$ to ARX$^\text{lo}$ state or whether such a transition itself is impaired.
While α cells enriched in lineage factors appear less mature and maintain some plasticity, whether T2D induces a true reversal of maturation or a more general phenotypic drift could still be questioned. While an enrichment of numerous gene sets including β cell genes in the ARX cells in T2D could suggest the latter, we note that those are also all enriched in true juvenile α cells (Avrahami et al., 2020). The exact links between these changes and dysregulated glucagon secretion remains somewhat speculative, particularly since we see no decrease in Ca2+ channel transcripts to explain the reduced Ca2+ currents. Interestingly, mitochondrial respiratory complex assembly transcripts are increased in T2D α cells, most notably in the α cells with impaired function and enriched for ARX. While altered mitochondrial function could drive P/Q-type Ca2+ channel inhibition and impaired exocytosis, and we provide some evidence linking the respiratory chain to α cell dysfunction in T2D, we do not know if mitochondrial respiration is altered in T2D α cells. This will require assessment of mitochondrial function and oxygen consumption in the ARX subset of α cells from donors with T2D as a future priority.

Finally, we should be careful when directly comparing the rodent and human studies. One clear difference we find is that human α cells in T2D show impaired (low) exocytosis at 1 mM glucose and a modest increase with higher glucose, while the mouse HFD α cells instead show altered Na+ channel inactivation and exocytosis-Ca2+-channel coupling. The exact reasons for these differences are unclear but perhaps related to obvious differences between humans and the mouse model (disease/donor pancreas for research. We also thank Dr. Rita Bottino (Allegheny Health Network) and Drs. James Shapiro and Tatsuya Kin (University of Alberta Clinical Islet Program) for contributing some islet preparations for this study. Finally, we especially thank the organ donors and their families for their kind gift in support of diabetes research. T.d.S. was supported by the Alberta-Helmholtz Diabetes Research School, the Alberta Innovates Scholarship in Data-Enabled Innovation, and the Sir Fredrik Banting and Dr. Charles Best Canada Graduate Scholarship. This work was funded by grants to P.E.M. from the Canadian Institutes of Health Research (CIHR; 148451); from the JDRF to P.E.M., L.B., and P.R. (SRA-2019-698-S-B); and from the National Institutes of Health (F32 DK109577 to E.M.W.; R01 DK126482 to R.S., P.E.M., and S.K.; and U01 DK120447 to P.E.M., R.A.D., and S.K.). Some islet samples were from the Human Pancreas Analysis Program (HPAP; RRID: SCR_016202), a Human Islet Research Network (RRID: SCR_014393) consortium (UC4-DK-112217 and UC4-DK-112232). P.E.M. holds the Canada Research Chair in Islet Biology.

AUTHOR CONTRIBUTIONS

X.-Q.D., J.C.-S., A.B., A.F.S., T.d.S., L.J.J.B., A.N., R.A.D., E.M.W., D.A., J.J., and R.C.J. collected and analyzed data. J.L., N.S., A.B., and J.E.M.F. isolated cells in T2D show impaired (low) exocytosis at 1 mM glucose and a modest increase with higher glucose, while the mouse HFD α cells instead show altered Na+ channel inactivation and exocytosis–Ca2+-channel coupling. The exact reasons for these differences are unclear but perhaps related to obvious differences between humans and the mouse model (disease/HFD duration, degree of dysglycemia, age, etc.). Nonetheless, in both mice and human cells, we find evidence to suggest that α cell dysfunction is linked to cell maturation state.

STAR+METHODS

Detailed methods are provided in the online version of this paper and include the following:

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SUPPLEMENTAL INFORMATION

Supplemental information can be found online at https://doi.org/10.1016/j.cmet.2021.12.021.

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The University of Alberta is situated on Treaty 6 territory, traditional lands of First Nations and Métis people. We thank Dr. Jesper Grud Skat Madsen (University of Southern Denmark), Dr. Jakob Knudsen (University of Copenhagen), and Dr. Lori Susel (University of Colorado) for helpful discussion and Dr. Francis Lynn (University of British Columbia) for critical reading of the draft manuscript. We thank the Human Organ Procurement and Exchange (HOPE) program and Trillium Gift of Life Network (TGLN) for their work in procuring human donor pancreas for research. We also thank Dr. Rita Bottino (Allegheny Health Network) and Drs. James Shapiro and Tatsuya Kin (University of Alberta Clinical Islet Program) for contributing some islet preparations for this study. Finally, we especially thank the organ donors and their families for their kind gift in support of diabetes research. T.d.S. was supported by the Alberta-Helmholtz Diabetes Research School, the Alberta Innovates Scholarship in Data-Enabled Innovation, and the Sir Fredrik Banting and Dr. Charles Best Canada Graduate Scholarship. This work was funded by grants to P.E.M.

DECLARATION OF INTERESTS

The authors declare no competing interests.

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REFERENCES


## KEY RESOURCES TABLE

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**Critical commercial assays**

| U-PLEX Mouse Glucagon Assay | Meso Scale Diagnostics | K1525YK |
| STELLUX Rodent Insulin Chemiluminescence ELISA | ALPCO | 80-INSMR-CH10 |
| KAPA HiFi HotStart ReadyMix | KAPA Biosystems | KK2601 |
| Nextera XT | Illumina | FC-131-1096 |

**Deposited data**

| Single cell mRNA-seq data | this paper | GEO: GSE164875 |
| Single cell mRNA-seq data | Camunas-Soler et al., 2020 | GEO: GSE124742 |
| Single cell mRNA-seq data | Avrahami et al., 2020 | GEO: GSE154126 |
| Processed patch-seq datasets | Camunas-Soler et al., 2020 | https://github.com/jcamunas/patchseq |

**Experimental models: Organisms/strains**

| mouse C57BL/6NCrl Inbred | Charles River Laboratories | CRL-027; RRID: IMSR_CRL:027 |
| 5L0D Picolab Laboratory Rodent Diet | LabDiet | 3005659-220 |
| Mouse Diet, High Fat Calories | VWR / Bio-Serv | CA89067-471 |

**Oligonucleotides**

| SmartSeq2 OligodT: 5’- AAGCAGTGGTATCAACGCAGAGT ACT30VN-3’ | Picelli et al., 2014 | N/A |
| SmartSeq2 TSO: 5’- AAGCAGTGGTATCAACGCAGAGT CATrGrG-3’ | Picelli et al., 2014 | N/A |
| SmartSeq2 ISPCR: 5’- AAGCAGTGGTATCAACGCAGAGT-3’ | Picelli et al., 2014 | N/A |

**Software and algorithms**

| Qlucore Omics Explorer v3.6 | Qlucore | https://www.qlucore.com/ |
| WEB-based Gene SeT AnaLysis Toolkit | Liao et al., 2019 | http://webgestalt.org/ |
| Graphpad Prism v9.0.0 | GraphPad | https://www.graphpad.com |
| Custom analysis software | Camunas-Soler et al., 2020 | https://github.com/jcamunas/patchseq |
| PatchMaster 2x90.1 | Smart Ephys HEKA | https://www.heka.com |
| FitMaster 2x90.1 | Smart Ephys HEKA | https://www.heka.com |
| STAR | Dobin et al., 2013 | https://github.com/alexdobin/STAR |
| HTSeq | Anders et al., 2015 | https://github.com/simon-anders/htseq |
| Qupath v0.2.3 | Bankhead et al., 2017 | https://qupath.github.io |
| Cytomap v1.4 | Stoltzfus et al., 2020 | https://gitlab.com/gernerlab/cytomap |
| Genomica | Segal Lab of Computational Biology | http://genomica.weizmann.ac.il/ |
| GSEA 4.1.0 | Subramanian et al., 2005 | http://software.broadinstitute.org/gsea/index.jsp |
| Python v3.7.11 | Python Software Foundation | http://www.python.org |
| XGBoost v1.0.2 | XGBoost | https://xgboost.readthedocs.io/en/latest/python/index.html |
| Scikit-learn v0.24.2 | Pedregosa et al., 2012 | https://scikit-learn.org/stable/ |
RESOURCES AVAILABILITY

Lead contact
Further information and requests for resources and reagents should be directed to and will be fulfilled by the lead contact, Patrick MacDonald (pmacdonald@ualberta.ca).

Materials availability
This study did not generate new unique reagents.

Data and code availability
Raw sequencing reads are available in the NCBI Gene Expression Omnibus (GEO) and Sequence Read Archive (SRA) under accession numbers GSE124742 and GSE164875.

The code and scripts generated during this study, as well as preprocessed datasets, are available at https://github.com/jcamunas/patchseq.

Data S1 represents an Excel file containing the values that were used to create all the graphs in the paper. Any additional information required to reanalyze the data reported in this paper is available from the lead contact upon request.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Human islets
In most cases, human islets were from our in-house human islet isolation and distribution program (https://www.isletcore.ca) (Lyon et al., 2019). Some human islets were provided by the Clinical Islet Transplant Program at the University of Alberta, by Dr. Rita Bottino at the Alleghany Health Network (Pennsylvania, US), or from the Human Pancreas Analysis Program (Kaestner et al., 2019). Details of donors with no diabetes (ND) or type 2 diabetes (T2D) used in this study are shown in Table S1. T2D was determined either by reporting of previous clinical diagnosis at the time of organ procurement, or by assessment of %HbA1c > 6.5 in a few cases that were considered as previously undiagnosed T2D. Human islets and dispersed cells were cultured in DMEM (ThermoFisher, #11885) with 10% FBS (ThermoFisher, #12483020) and 100 U/ml penicillin/streptomycin (Thermo Fisher, #15070063) at 37°C and 5% CO2. In our previous study (Camunas-Soler et al., 2020) we saw no obvious effect of sex on patch-seq data, or when analyses were corrected for sex as a co-variate. In the present study we find that electrophysiological fingerprint modelling is not impacted by donor sex (Figure 6B). All donors provided written informed consent for research. Human tissue studies were approved by the Human Research Ethics Board of the University of Alberta (Pro00013094, Pro00001754).

Mouse islets
Mouse islets were isolated from chow-fed (5L0D Picolab Laboratory Rodent Diet, #3005659-220) male C57bl/6N mice (Charles River, #CRL:027; RRID: IMSR_CRL:027) at 10-12 weeks of age or following 10-12 weeks of high fat diet (60% of calories from fat; VWR Bio-Serv, #CA89067-471) starting from 8 weeks of age by collagenase digestion and hand-picking (Smith et al., 2020). Mouse islets and dispersed cells were cultured in RPMI (ThermoFisher, #11875) with 10% FBS and 100 U/ml penicillin/streptomycin at 37°C and 5% CO2. Animals were housed at 20-24°C with a 12h:12h light:dark cycle and daily health checks. Studies were performed in accordance with institutional guidelines and were approved by the Animal Care and Use Committee at the University of Alberta (AUP00000291).

METHOD DETAILS

Patch-clamp recordings
Hand-picked islets were dissociated to single cells using StemPro accutase (Thermo Fisher, #A1110501). Dispersed islet cells were cultured for 1–3 days, after which media was changed to bath solution containing (in mM): 118 NaCl, 20 TEA, 5.6 KCl, 1.2 MgCl2, 2.6 CaCl2, 5 HEPES, and with glucose as indicated (pH adjusted to 7.4 with NaOH) in a heated chamber (32–35°C). Modulators of Ca2+ channels (isradipine, Sigma-Aldrich, #I6658; agatoxin, Alomone Labs, #STA-500; GV-58, Alomone Labs, #G-140) or mitochondrial respiratory chain inhibitors (antimycin A, Sigma-Aldrich, #A8674; rotenone, Sigma-Aldrich, #R8875) were added to the bath, and pH adjusted, as indicated in figure legends. For whole-cell patch-clamping, fire polished thin wall borosilicate pipettes coated with Sylgard (3–5 MOhm) contained intracellular solution (in mM): 125 Cs-glutamate, 10 CsCl, 10 NaCl, 1 MgCl2, 0.05 EGTA, 5 HEPES, 0.1 cAMP and 3 MgATP (pH adjusted to 7.15 with CsOH). Electrophysiological measurements were collected using a HEKA EPC10 amplifier and PatchMaster Software (SmartEphys HEKA) within 5 minutes of break-in as described previously (Camunas-Soler et al., 2020). Quality control was assessed stability of the seal (>10 GOhm) and access resistance. Cells were identified by post-hoc immunostaining for insulin with a rabbit anti-insulin primary antibody (Santa Cruz; #SC-9168; RRID: AB_2126540) and goat anti-rabbit Alexa Fluor 488 secondary (ThermoFisher, #A-11076; RRID: AB_141930), and with a guinea pig anti-glucagon primary antibody (Sigma-Aldrich, #G2654; RRID: AB_259852) and goat anti-guinea pig Alexa Fluor 594 secondary (ThermoFisher, #A-11076; RRID: AB_141930); or following collection for single-cell RNA sequencing analysis (Camunas-Soler et al., 2020).
For assessment of Ca\(^{2+}\) channel activity using Ba\(^{2+}\) as a charge carrier (Figures 1C and 1F) bath solution contained (in mM): 100 NaCl, 20 BaCl\(_2\), 5 CsCl, 1 MgCl\(_2\), 10 HEPES, with 0.2 \(\mu\)M tetrodotoxin and glucose as indicated (pH 7.4 with NaOH); and intracellular solution was (in mM): 140 Cs-glutamate, 1 MgCl\(_2\), 20 TEA, 5 EGTA, 20 HEPES, 3 Mg-ATP (pH 7.15 with CsOH). Compounds used were: lysophosphatidic acid (LPA; Sigma-Aldrich, #L7260), \(\alpha\)-latrotoxin (Enzo Life Sciences, #ALX-630-027-C040), Slit guidance ligand 2 (SLIT-2-N; Sigma-Aldrich, #SRP31555), prostaglandin E\(_2\) (PGE\(_2\); Tocris, #2296), somatostatin (SST; Sigma-Aldrich, #S9129), glucose-dependent insulinotropic polypeptide (GIP; Eurogentec, #AS-65568), L-glutamic acid (Sigma-Aldrich, #G1251), and epinephrine (Sigma-Aldrich, #E4375). These were added in bath solution, at the concentrations indicated in the figure legend, and pH was re-adjusted if needed. Na\(^{+}\) current was activated by a 50 ms depolarization from -70 to -10 mV.

Pancreas patch-seq

Our protocol for pancreas patch-seq is outlined in a recent paper (Camunas-Soler et al., 2020). In brief, following patch-clamp cells were collected using a separate wide-bore collection pipette (0.2-0.5 MOhm) filled with lysis buffer (10% Triton, Sigma-Aldrich, #93443), ribonuclease inhibitor 1:40, Clontech, #2313A), ERCC RNA spike-in mix (1:60000; ThermoFisher, #4456740), 10 mM dNTP, and 100 \(\mu\)M dT (3'-AAGCAGTGGTATCAACGCAGAGTACTTTTTTTTTTTTTTTTTTTTTTTTTTTTVN-5') and then transferred to PCR tubes and stored at -80°C. We generated cDNA and sequencing libraries using an adaptation of the SmartSeq-2 protocol for patch-seq plates (Camunas-Soler et al., 2020; Picelli et al., 2014). Libraries were generated from the amplified cDNA by tagmentation with Tn5 and sequenced in the NovaSeq platform (Illumina) using paired-end reads (100 bp) to an average depth of 1 million reads per cell. Sequencing reads were aligned to the human genome (GRCh38 genome with supplementary ERCC sequences) using STAR (Dobin et al., 2013), and gene counts determined using htseq-count (intersection-nonempty) using a GTF annotation with Ensembl cell. Raw sequencing reads are available at the NCBI Gene Expression Omnibus (GEO) and Sequence Read Archive (SRA) under accession numbers GSE124742 and GSE164875. Correlation between transcript expression and electrophysiology was as outlined previously (Camunas-Soler et al., 2020). Visualization as t-distributed stochastic neighbor embedding (tSNE) plots was with the Qlucore Omics Explorer v3.6. Gene-set enrichment analysis (GSEA) was performed with of Z-scores or slopes for transcripts found in >20% of cells using the WEB-based Gene SeT AnaLysis Toolkit (http://webgestalt.org/) and weighted set gene coverage to reduce redundancy in identified terms (Liao et al., 2019).

Hormone secretion measurements

Groups of 150-175 hand-picked mouse islets were perfused using a BioRep perfusion system (BioRep, Miami). Islets were pre-incubated in perfused KRB buffer containing (in mM): 140 NaCl, 3.6 KCl, 2.6 CaCl\(_2\), 0.5 NaH\(_2\)PO\(_4\), 0.5 MgSO\(_4\), 5 HEPES, 2 NaHCO\(_3\) and 0.5 mg/ml Essentially fatty acid free BSA (Sigma A6003) for 30 minutes, and then perfused in the same KRB with changes in glucose and KCl as indicated. Samples were collected at 90 -210 sec intervals and stored for assay of glucagon (U-PLEX Mouse Glucagon Assay, Meso Scale Diagnostics, #K1525YK) and insulin (STELLUX Rodent Insulin Chemiluminescence ELISA, ALPCO, #B0-INSMR-CH10) at -20°C.

Immunostaining and single-cell protein analysis

Human paraffin embedded pancreas biopsies were sectioned to 3 \(\mu\)m and immunostained with antibodies against MAFB (Cell Signaling Technologies, #41019; RRID:AB_2799192), ARX (R&D Systems, # AF7068; RRID:AB_10973178) and GCG (Sigma-Aldrich, #G2654; RRID:AB_259852). The secondary antibodies were Cy2, Cy3, or Cy5 conjugated (1:50; Jackson ImmunoResearch, 115-165-003; RRID:AB_2338680, 713-175-147, RRID:AB_2340730). Nuclear co-staining was conducted with DAPI Fluoromount G (Southern Biotech, #010-20). Immunofluorescent images were acquired on a Zeiss Axio Imager M2 widefield microscope with ApoTome. Images containing islet and exocrine cells were processed for nuclei segmentation, individual cell detection and cytosol border inference using Qupath software v0.2.3 (Bankhead et al., 2017). Parameters for nuclei detection: background radius 20px, median filter 5px, sigma 7px, minimum nuclei area 10px\(^2\), maximum nuclei area 1000px\(^2\), threshold of 2, nuclei were split by shape, and cell boundaries were determined by an expansion threshold of 12px with smoothing. Next, the relative expression levels of nuclear ARX, nuclear MAFB and cytosolic GCG levels for each detected cell were exported as.csv files and imported into CytoMap v1.4 for spatial analysis (Stoltzfus et al., 2020). In Qlucore Omics Explorer v3.6, we performed dimensionality reduction using tSNE of all imaged islet dataset using default parameters with 30 perplexity and 0.5 theta. Data was normalized for each dataset and the relative GCG, ARX and MAFB levels were displayed in the tSNE plot. Next, we created a gate to select the \(\alpha\)-cell population in our dataset and the relative levels of ARX and MAFB were plotted. This information was used to determine the identity of ARX\(^{hi}\) and ARX\(^{low}\) cells and the relative expression levels of ARX, MAFB and GCG for each \(\alpha\)-cell subpopulation was plotted.
Gene-set and pathway analysis of scRNA-seq
To characterize the expression program of ARXhi and ARXlow α-cells in ND and T2D samples, we conducted gene set enrichment analysis (GSEA) on aggregated ARXhi versus ARXlow α-cell transcriptomes from our recently published data (Avrahami et al., 2020) using a hypergeometric test by Genomica software (http://genomica.weizmann.ac.il/) considering gene sets with a p value < 0.01 and an FDR <0.05 to be significantly enriched. Pathway analyses were performed with Gene Set Enrichment Analysis (GSEA) (Subramanian et al., 2005).

QUANTIFICATION AND STATISTICAL ANALYSIS

Electrophysiological fingerprint Modeling
Multiple regression was carried out on ND α-cells using Ordinary Least Squared (OLS) Regression with Statsmodel v0.12.2 (Seabold and Perktold, 2010). Independent variables included age, sex, body mass index (BMI), HbA1c, cold ischemia time (CIT), culture time, and glucose concentration. Dependent variables included: cell size (pF), total exocytosis (fF/pF), early exocytosis (fF/pF), late exocytosis (fF/pF), peak Na+ current (pA/pF), Na+ half-Inactivation (mV), early Ca2+ current (pA/pF), late Ca2+ current (pA/pF), Ca2+ charge entry during an initial depolarization (pC/pF), exocytosis normalized to Ca2+ charge entry (fF/pC), reversal potential (mV), and α probability from Models 1-3 (see below). Cells lacking data for a dependent variable were only dropped from that specific OLS analysis. Classification of cell type was conducted using the above electrophysiological measures as dependent variables from ND donors in an Optimizable Ensemble that either included (Model 1) or excluded (Model 2) cell size in MATLAB, or using Extreme Gradient Boosting (XGBoost v1.0.2) in a Python v3.7.11 framework that excluded cell size and reversal potential and restricted training data to 32 ND donors within an age of 20-70 years, a BMI of 18.5-30.3, and a pancreas CIT of ≤20 hours. Fine tuning was performed with a pre-determined minimum accuracy of 80% for both α- and β-cells and training was performed with early stopping set to 50 iterations and utilized AUCPR as the evaluation metric. Models were trained on 80% of the α- and β-cells, with 20% reserved for testing and validation. Confusion matrices were generated using scikit-learn v0.24.2 (Pedregosa et al., 2012). We applied the models across our combined immunostaining and patch-seq database of ND and T2D cells and utilized the classifier’s predicted probability scores to assess fit to α-cell (αprobability = 1.0) and β-cell (αprobability = 0.0) models.

Statistical analysis
Data are expressed as mean and standard error (line plots), mean and 10-90 percentile range (box and whisker plots), or as violin plots with median and quartiles indicated. When comparing two groups we used Student’s t test or the non-parametric Mann-Whitney test to compare ranks. When comparing more than two groups we used one-way or two-way ANOVA followed by either the Tukey post-test or two-stage step-up method for estimation of false discovery rate (FDR); or alternatively the non-parametric Kruskal-Wallis test followed by Dunn’s post-test to correct for multiple comparisons. Statistical tests used are indicated in the figure legends. p values less than 0.05 were considered as significant. For correlation of electrophysiology with transcript expression, Spearman tie-corrected correlations were computed for each gene and significance was tested by bootstrapping (1,000 iterations) as described in our previous work (Camunas-Soler et al., 2020). For ORA and GSEA, false discovery rate (FDR) of reported pathways was <0.05 or <0.1 as indicated in the figures using Bengamini-Hochberg (BH) correction within the WEB-based GEn e SeT AnaLysis Toolkit (webgestalt.org). We did not perform power calculations prior to experiments, and experimenters were not blinded. Animals were assigned randomly to CD and HFD groups, and all human donors were accepted for islet isolations if they met the requirement for negative serology reports. No cells that passed standard electrophysiology and sequencing quality control (described above) were excluded from analysis.