



Richard K.P. Benninger^{1,2} and David J. Hodson^{3,4,5}

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Insulin-secreting β -cells are heterogeneous in their regulation of hormone release. While long known, recent technological advances and new markers have allowed the identification of novel subpopulations, improving our understanding of the molecular basis for heterogeneity. This includes specific subpopulations with distinct functional characteristics, developmental programs, abilities to proliferate in response to metabolic or developmental cues, and resistance to immune-mediated damage. Importantly, these subpopulations change in disease or aging, including in human disease. Although discovering new β-cell subpopulations has substantially advanced our understanding of islet biology, a point of caution is that these characteristics have often necessarily been identified in single β -cells dissociated from the islet. β -Cells in the islet show extensive communication with each other via gap junctions and with other cell types via diffusible chemical messengers. As such, how these different subpopulations contribute to in situ islet function, including during plasticity, is not well understood. We will discuss recent findings revealing functional β -cell subpopulations in the intact islet, the underlying basis for these identified subpopulations, and how these subpopulations may influence in situ islet function. Furthermore, we will discuss the outlook for emerging technologies to gain further insight into the role of subpopulations in in situ islet function.

In most vertebrate species, β -cells form clusters with other hormone-secreting cells (glucagon-secreting α -cells, somatostatin-secreting δ -cells) within islets of Langerhans. Very early studies of the β -cell assumed them to be homogenous based on a lack of morphological differences. However, detailed studies subsequently determined that there exists a broad heterogeneity in the function of β -cells. These early studies of β -cell heterogeneity are summarized by the landmark review of Pipeleers (1), which describes with remarkable foresight the presence, characteristics, and role of functional β -cell subpopulations. This includes how dissociated β -cells show functional heterogeneity, with populations of cells displaying higher levels of glucose metabolism, redox state, insulin synthesis, membrane potential, and insulin secretion; that morphological markers (nuclear size, insulin granularity) can differentiate β -cell subpopulations with differing glucose sensitivity and insulin secretion levels; that β -cells show heterogeneous expression of key proteins such as glucokinase (GCK), connexins, or insulin, including spatial variations across the islet; that β -cells with low glucosestimulated insulin secretion preferentially increase in number under development or metabolic stress; and that β -cells vary in their sensitivity to cytotoxic agents. Despite this indepth knowledge, there have been several gaps in our understanding that have persisted until recently:

- 1. What is the molecular basis for β -cell functional diversity?
- 2. Which markers can be used to identify and characterize β -cell subpopulations?
- 3. Does functional heterogeneity in the intact islet or pancreas mirror that observed among dissociated β-cells?

A β -cell is a terminally differentiated cell that produces and secretes insulin in a glucose-regulated manner. Importantly, β -cells have the ability to adapt to changes in metabolic demand through increased insulin secretion and/or number.

- ¹Department of Bioengineering, University of Colorado Anschutz Medical Campus, Aurora, CO
- ²Barbara Davis Center for Diabetes, University of Colorado Anschutz Medical Campus, Aurora, CO
- ³Institute of Metabolism and Systems Research, University of Birmingham, Birmingham, U.K.
- ⁴Centre for Endocrinology, Diabetes and Metabolism, Birmingham Health Partners, Birmingham, U.K.
- ⁵Centre of Membrane Proteins and Receptors (COMPARE), University of Birmingham and University of Nottingham, Birmingham, U.K.

Corresponding authors: Richard K.P. Benninger, richard.benninger@ucdenver .edu, and David J. Hodson, d.hodson@bham.ac.uk.

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- 4. What is the role of β-cell heterogeneity in islet function and glucose homeostasis, and can changes in heterogeneity contribute to diabetes?
- 5. Are β -cells fixed in specific functional states, or can they transition between states over time?

We will describe recent technological advances and studies that have answered some of these key questions, with a focus on understanding the consequence of heterogeneity in β -cell function within the islet setting.

RECENT ADVANCES CHARACTERIZING $\beta\text{-CELL}$ HETEROGENEITY

Early and more recent studies demonstrated heterogeneity in insulin secretion in dissociated mouse or human β -cells using the hemolytic plaque assay (2). Patch-clamp measurements also revealed heterogeneity in dissociated β -cell electrical properties (3). Autofluorescence measurements revealed heterogeneity in redox state, and incorporation of radioactive tracers revealed heterogeneity in glucose metabolism and insulin biosynthesis (4). The development of fluorescent biosensors and confocal or 2-photon microscopy provided tools to further characterize β -cell functional differences. This includes precise quantification of heterogeneity in dissociated β -cell glucose metabolism and redox state (5); glucose sensitivity to Ca²⁺ elevations and Ca²⁺ oscillation patterns (6); and cAMP oscillation patterns (7). Recently, the application of new biomarkers or high-throughput single-cell analyses has further revealed molecular details underlying β-cell heterogeneity.

Markers of β -Cell Subpopulations

Early studies suggested insulin granularity was a morphological marker that could separate a population of β -cells with a low glucose threshold (4). More recently, several markers have been used that reveal β -cell subpopulations with differing function. Polysialylated-neural cell adhesion molecule (PSA-NCAM) separated two populations of mouse β-cells, with one population (β^{high}) showing higher Ca²⁺ and ATP elevation, insulin secretion, and Gck and Glut2 expression (8). Insulin promoter activity (MIP-GFP fluorescence) separated three populations of β -cells, with the MIP-GFP^{low} population (~10% incidence in adult) possessing low insulin expression and low granularity (9). Aguayo-Mazzucato et al. (10) subsequently showed that the MIP-GFP^{low} and MIP-GFP^{high} populations decreased and increased in incidence, respectively, during aging. Further, the MIP-GFP^{low} population was not marked by increases in the aging markers IGF1R or p16Ink4a, the lack thereof indicative of newly formed cells. The authors also showed that a population of highly secreting cells is lost during aging. However, the low incidence of this population (\sim 5%) is suggestive of a subset of MIP-GFP^{high} cells. Recently, Lickert and colleagues (11) used a Fltp-Venus reporter mouse to separate two populations of β -cells. Fltp-Venus positive (Fltp⁺) cells showed improved insulin secretion but reduced proliferative ability, together with more extensive mitochondrial morphology,

more oxidative metabolism, and increased insulin granularity. The genetic profile of Fltp⁺ cells was also consistent with improved function (i.e., increased expression of genes for GCK, GLUT2, GLP1R, GIPR, and Cx36), as well as increased maturity. Notably, during development or upon metabolic stress the Fltp⁻ population showed greater proliferation and number. These studies together indicate a dynamic β -cell subpopulation defined by improved glucose-stimulated insulin secretion as a result of elevated glucose metabolism and insulin granule content.

Huising and colleagues (12) used an Ucn3 reporter mouse to reveal a population of insulin⁺ Ucn3⁻ cells undergoing α -cell-to- β -cell transition at the intact islet periphery. These immature cells showed decreased expression of genes regulating glucose metabolism (Gck, tricarboxylic acid cycle, oxidative phosphorylation), membrane potential (K_{ATP}, Ca_V subunits), and *Glp1r* and *Ins*. They also lacked both glucose influx and glucose-regulated membrane potential and Ca²⁺ influx. Although reminiscent of subpopulations with reduced function, this population is much lower in incidence $(\sim 2\%)$, shows negligible glucose sensitivity, and is not differentiated by markers such as Fltp. As such, this represents a novel cell subpopulation. Herold and colleagues (13) also identified a population of β -cells in NOD mice characterized by reduced granularity that increased in number during autoimmune attack of the islet. This population also displayed reduced levels of insulin, GLUT2, and maturity markers, consistent with earlier studies (1). Thus, rarer subpopulations of β -cells do exist, although how they overlap is not known.

In human β -cells, Grompe and coworkers (14) identified two surface markers (CD9, ST8SIA1) that separated four populations of β -cells with distinct gene expression profiles. ST8SIA1⁺ cells (~15% in healthy donors) showed lower insulin secretion and lower expression of *GLUT2* but increased expression of genes encoding K_{ATP} channel subunits and other K⁺ channels. The incidence of ST8SIA1⁺ cells is also increased in donors with type 2 diabetes.

Although these markers all separate β -cell populations with distinct functional states, further less well-characterized subpopulations have been identified using other markers (see Roscioni et al. [15]). However, it is important to remember that markers are not generally associated with driving the functional state of the β -cell subpopulation. For example, a Fltp deletion has minor impact on β -cell function (11). Thus, PSA-NCAM or CD9/ST8SIA1 may simply separate differing populations in an arbitrary manner. Furthermore, it is not always clear whether fluorescent reporter constructs reflect the protein level. For example, MIP-GFP^{medium} and MIP-GFP^{high} β-cells show identical insulin transcription and protein levels (9), and the change in *Fltp* gene transcription only varies \sim twofold between Fltp⁺ and Fltp⁻ cells defined by the Fltp-Venus reporter (11). Thus, great care should be taken when linking the marker involved to the characteristics of the subpopulation in question.

High-Throughput Single-Cell Analysis

In an elegant study, Kaestner and colleagues (16) used mass cytometry to detect large numbers of protein markers in human β -cells, indicating three consistent β -cell states (C1, C2, C3). One state (C1, 10–70% in adults) showed substantially lower proliferation than the other two states (C2, C3), where the proportion of C1 increased with age and decreased in obesity. Although limited probes for proteins involved in β -cell function were used, the C1 population showed higher PDX1 and insulin expression indicative of a mature functional population. The C1 population also showed lower ST8SIA1 and CD9 levels compared with proliferating C2, C3 populations, where previously ST8SIA1⁻ and CD9⁻ β -cells showed improved insulin secretion and maturity markers and decreased in proportion during type 2 diabetes (14).

Single-cell sequencing has been used by several groups to examine the transcriptional profile in islet endocrine cells (17,18) (for a comprehensive review on this topic, see Gutierrez et al. [19]). This has also resolved subpopulations of β -cells in humans (17), identifying five β -cell states. Interestingly, all showed similar insulin expression but differed in gene transcription that included free fatty acid receptor 4 and RBP4 (an adipokine receptor). However, it remains unclear whether these states overlap with those revealed by mass cytometry or markers described to date. Further, no clear heterogeneity was detectable using similar analyses in mouse β -cells (20), questioning whether this may reflect true species differences, a reduction in variability due to use of syngeneic donors maintained in controlled housing conditions, or an effect of isolation protocol. Moreover, single-cell analysis characterizes heterogeneity in dissociated β -cells, and islet dissociation may bias the population of cells that are analyzed, for example, due to reduced β-cell viability. Integration of multiple factors (e.g., environmental stimuli, developmental cues, cell cycle and tissue architecture) that affect cell identity will also be important in future studies to robustly define subpopulations (21).

Interpreting Subpopulations From Marker and Single-Cell Analyses

Figure 1 summarizes the functional β -cell subpopulations identified by recent marker and single-cell sequencing studies and their relative overlap in characteristics. Bearing in mind potential limitations in marker studies or highthroughput single-cell studies, these studies consistently show a mature population of β -cells (incidence 50–80%) that is functionally competent, probably due to higher rates of oxidative metabolism and ATP production; high levels of insulin expression, granule content, and secretion; and potentially improved electrical properties. This population reduces in incidence or shows dysfunction under aging and diabetes, possibly due to increased susceptibility to endoplasmic reticulum stress (22). Conversely, the opposing less mature population (incidence 15–50%) shows reduced insulin secretion secondary to reductions in glucose metabolism and insulin expression but has improved proliferative capacity.

Although some studies indicate that more than two populations exist (e.g., MIP-GFP $^{\rm low/medium/high},\, {\rm CD9}^{+/-}/$ $ST8SIA1^{+/-}$), it is unclear whether the "functionally competent" and "less functional" populations can be further subdivided. Likely, there exist small distinct subpopulations, such as the Ucn3⁻insulin⁺ neogenic niche, with some similarities to the less functional population. Future work will require caution when determining whether identified subpopulations truly are distinct or whether they are an artifact of the analysis used. For example, does increasing the number of markers generate more or less apparent/ overlapping β -cell populations, or can populations be reproduced in a second data set (e.g., using training and test sets)? Can the absence of a subpopulation be excluded given the relatively poor resolution of 'omics approaches such as RNA-Seq, where even at 6,000 M reads a proportion of genes remains unquantified?

Importantly, most studies to date described above have characterized subpopulations using single-cell preparations that lack the influence of the interactions between β -cells and other cell types within the islet. Although reaggregation methods can recover some of these influences, it is still unclear how different subpopulations may interact within the islet to influence islet function. Therefore, proper consideration of how a heterogeneous population of β -cells functions within the islet is needed.

HOW HETEROGENEOUS CELLS INTERACT IN THE ISLET

Interactions between cells within the islet are critical: dissociation of the islet into single β -cells reduces the dynamic range of glucose-stimulated insulin secretion by \sim 10-fold (23). Even when β -cells remain coupled to one or two other β - or α -cells, insulin secretion significantly improves (24). Although many mechanisms of cell–cell communication have been identified, less is known about how they coordinate heterogeneous cell populations.

Gap Junction Electrical Coupling Between Heterogeneous Cells

Gap junction (GJ) channels electrically couple β -cells within mouse and human islets (25), serving two main functions. First, GJ channels coordinate oscillatory dynamics in electrical activity and Ca²⁺ under elevated glucose or GLP-1, allowing pulsatile insulin secretion (26,27). Second, GJ channels lower spontaneous elevations in Ca²⁺ under low glucose levels (28). GJ coupling is also heterogeneous within the islet (29), leading to some β -cells being highly coupled and others showing negligible coupling.

Several studies have examined how electrically heterogeneous cells interact via GJ channels using K_{ATP} channel mutant mice (Fig. 2). The Kir6.2^[Δ N30,K185Q] mouse model CreER-dependently expresses overactive K_{ATP} channels in a population of β -cells, rendering them inexcitable (30). By varying recombination efficiency, as few as ~15% inexcitable Α

Functional subpopulations									
Population	Characteristics	Gene and protein expression	Incidence	Proliferation	Maturity	Plasticity			
Mip- GFP ^{high/medium}	High insulin granule content	High insulin expression	~90%(i)	Increased senescence	mature	Increased with age			
PSA-NCAM ⁺ (β ^{high})	High [Ca ²⁺], ATP, cAMP elevation, improved insulin secretion	High <i>Gck</i> , <i>Glut2</i> expression. High exocytotic machinery levels.	~50%(ii)	A.	mature	Decrease in obesity			
Fltp⁺	Increased oxidative metabolism, increased insulin granule content	High expression of genes for GCK, GLUT2, GLP1R, GIPR, Cx36, cellular respiration, mitochondrial organization	~80%	Lower proliferative capacity	mature	Decrease during development, metabolic stress,			
ST8SIA1 ⁻	Improved insulin secretion	High expression of GLUT2	~85% ⁽ⁱⁱⁱ⁾	-	mature	Reduced in T2D			
C1 (cytof)	-	High insulin expression	10-70% ^(iv)	-	higher Pdx1	Increased with age, decrease in obesity.			
Hub	High [Ca ²⁺] recruitment,	High GCK, low insulin levels	1-10%	-	Relatively immature	Disrupted by diabetogenic environment			
ChR2 ^{high}	High [Ca ²⁺] recruitment, high NAD(P)H	-	~5%	<u>-</u>	-	Disrupted by diabetogenic environment			

Nonfunctional subpopulations								
Population	Characteristics	Gene and protein expression	Incidence	Proliferation	Maturity	Plasticity		
Mip-GFP ^{low}	Low insulin granule content	Low insulin expression	~10% ⁽ⁱ⁾	No senescence with age	immature	Reduced with age		
PSA-NCAM [−] (β ^{low})	Low[Ca ²⁺] elevation, ATP elevation, reduced insulin secretion	Low Gck, Glut2 expression	~50%(ii)	-	Immature	Increase in obesity		
Fltp ⁻	Reduced oxidative metabolism, reduced insulin granule content	Low expression of genes for GCK, GLUT2, GLP1R, GIPR, Cx36, cellular respiration, mitochondrial organization	~20%	Higher proliferation	immature	Increased during development, metabolic stress,		
ST8SIA1+	Reduced insulin secretion	Low expression of GLUT2, higher expression K+ channels	~15%(iii)		immature	Increased in T2D		
Ucn3 ⁻	No glucose influx, membrane depolarization or [Ca ²⁺] elevation	Decreased expression of genes for GCK, TCA cycle, ox-phos; K_{ATP} , Ca _V , GLP1R, Insulin.	~2-3%(v)	α-β transition	very immature	Increased in development		
'bottom'	Low insulin granule content. Resistant to immune damage	Low GLUT2, insulin expression	~15%	Higher proliferation	Immature	Increased with T1D progression		
ChR2 ^{low}	Low [Ca ²⁺] recruitment, low NAD(P)H, wave initiation	-	~10%		-	Disrupted by diabetogenic environment		

C MipGFPhigh, PS-NCAM⁺, Fltp⁺, ChR2^{high}, Hub



MipGFPlow, PS-NCAM⁻, Fltp^{-,} Ucn3⁻



Figure 1—Summary of previously identified β -cell subpopulations using biomarker or single-cell analyses. *A*: Summary of functionally competent (functional) cell subpopulations, including their functional and gene/protein expression profile, incidence, and how they change in conditions associated with diabetes. *B*: Summary of less functional (nonfunctional) cell subpopulations. For *A* and *B*, shaded rows are functionally defined via optogenetics in situ. C: Schematic suggesting qualitative overlap between several identified subpopulations. Dashed lines indicate those defined via optogenetics in situ. Note the strong overlap between most identified larger populations, but the low-incidence functional hub cell subpopulation and nonfunctional Ucn3⁻ subpopulation show relatively lower overlap. This may reflect the number and nature of the markers measured, as well as the functional readout. ox-phos, oxidative phosphorylation; T1D, type 1 diabetes; TCA, tricarboxylic acid. (i) GFP^{high} increases from 10–40% with age (from 5–9 weeks to 16–40 weeks) and GFP^{low} decreases from 50 to 10% with age. (ii) PSA-NCAM⁺ is defined as top 50% expression. (iii) Incidence was 50% in donors with type 2 diabetes (T2D). (iv) Incidence was <30% in donors with T2D. (v) Ucn3⁻, insulin⁺ is located at the islet periphery. Adapted from Servier Medical Art under a CC-BY3.0 license (https://creativecommons.org/ licenses/by/3.0/).



Figure 2—Example for how heterogeneous cells can interact. *A*: Introducing Kir6.2^[Δ N30,K1850] into β -cells renders them unresponsive to glucose. With a small (~10%) population of these less functional cells in the islet (dark green), glucose responsiveness is unchanged. However, with ~20% of these cells, glucose responsiveness is lost. Glucose responsiveness is recovered by electrically isolating (Cx36ko) these less functional cells. *B*: Conversely, if a large proportion (~70%) of hyperexcitable Kir6.2^[Δ AA] cells are introduced into the islet (light green), glucose responsiveness remains, although this is lost when electrical coupling is lost (Cx36ko). *C*: In the native islet, the number of intrinsically activated (light gray) cells increases as glucose concentration rises. However, only when a certain threshold of cells are activated does Ca²⁺ elevate as a result of insufficient numbers of nonactivated cells (dark gray), analogous to *A* and *B*. *D*: The presence of activated and nonactivated cells can be further revealed by the K_{ATP} opener diazoxide, which renders only a small number of cells intrinsically inactive as observed in the absence of GJ coupling but fully abolishes Ca²⁺ elevations in the presence of GJ coupling. Adapted from Servier Medical Art under a CC-BY3.0 license (https:// creativecommons.org/licenses/by/3.0/).

 β -cells were found to be sufficient to suppress Ca²⁺ responses to glucose across the islet (30). Thus, small numbers of inexcitable cells can cause marked dysfunction by suppressing islet electrical activity and insulin secretion. Similar conclusions were made using the Kir6.2^[AAA] mouse model that expresses inactive K_{ATP} channels in ${\sim}70\%$ of β -cells, rendering them hyperexcitable (31). Despite the majority of β -cells lacking glucose-regulated K_{ATP} closure, islets from these mice showed near-normal responses, as the remaining ${\sim}30\%$ of normal $\beta\text{-cells}$ were sufficient to regulate islet Ca2+ (31). In each case, a deficiency in GJ channels led to KATP mutant cells and normal cells behaving distinctly (30,31). Thus, GJ coupling is a major mechanism by which electrically heterogeneous cells interact. Early studies imposing metabolic heterogeneity through a knockout of GCK in a population of β -cells did not find evidence of significant metabolic coupling between β -cells (5).

This series of experiments indicate a "bistability" in islet function, where a threshold number of poorly responsive β -cells is sufficient to totally suppress islet function. Notably, when islets lacking GJ channels are treated with low levels of the K_{ATP} activator diazoxide or the GCK inhibitor mannoheptulose, a subpopulation of cells are silenced, presumably corresponding to the less functional population (30). Only diazoxide/mannoheptulose concentrations capable of silencing >40% of these cells will fully suppress Ca²⁺ elevations in normal islets. Again, this indicates that a threshold number of poorly responsive cells can inhibit the whole islet. Thus, if there exists a threshold number of functionally competent β -cells (~60–85%), then the islet will show coordinated elevations in Ca²⁺ and insulin secretion.

Below this threshold number, the islet will lack Ca^{2+} elevation and insulin secretion (Fig. 2). The precise threshold depends on the characteristics of the excitable and inexcitable populations: small numbers of inexcitable cells will increase the number of functionally competent cells required for islet activity, whereas small numbers of highly excitable cells will do the opposite. However, if GJ coupling is lowered, then inexcitable cells will exert a reduced suppression, also decreasing the threshold required.

Paracrine and Juxtacrine Communication

Paracrine communication between β -cells and other endocrine cells is also important for regulating insulin secretion. Complex cross talk between α - and β -cells—stemming from neurotransmitters, ions, and hormones—shapes glucagon and insulin secretion depending on glucose concentration (reviewed in Gaisano et al. [32]). δ -Cell somatostatin secretion also inhibits insulin secretion. Notably, Ucn3 cosecreted with insulin promotes somatostatin secretion, leading to negative feedback against overt fluctuations in insulin secretion (33). Therefore, dynamic intercommunication exists between the main endocrine cells types within the islet. Several neurotransmitters (e.g., acetylcholine, glutamate, GABA) are also released within the islet by nerve terminals and endocrine cells, which can modulate insulin secretion and act on other endocrine cells (reviewed in Caicedo [34]). Juxtacrine- or contact-dependent communication includes EphA-ephrinA bidirectional signaling between β -cells (35). EphA forward signaling predominates at elevated glucose and promotes insulin secretion, but ephrinA reverse signaling predominates at low glucose and suppresses insulin secretion (35). NCAM signaling between β -cells also promotes insulin secretion (8).

Little is known how these paracrine and juxtacrine mechanisms impact heterogeneous cells. One possibility is a homogenizing effect. For example, given heterogeneity, some β -cells will secrete more insulin and cosecrete more Ucn3, thus stimulating more local somatostatin secretion. As a result, there may be a greater local suppression of insulin secretion, creating more homogeneous release. Whether such a homogenizing effect occurs with the various paracrine communication feedback loops (e.g., somatostain–Ucn3, glucagon–insulin) remains to be determined. Similarly, juxtacrine mechanisms may impact heterogeneous cells differently, particularly give heterogeneity in β -cell glucose metabolism. However, these actions also remain to be determined.

HETEROGENEITY IN THE INTACT ISLET

Functional heterogeneity in the islet in terms of glucose metabolism (NADH response, GCK expression) (5,36) has been long known. Although electrical activity and cAMP responses are apparently homogeneous in the islet (26,37), this is likely due to GJ coupling. In the absence of GJ coupling, the electrical response is heterogeneous in terms of glucose sensitivities and patterns of Ca²⁺ oscillations (26). Despite several aspects of function being homogeneous in the islet, β -cell subpopulations can dramatically influence overall islet function as discussed above.

A Role for Existing Subpopulations?

MIP-GFP^{high}, PSA-NCAM^{high}, Fltp⁺, and ST8SIA1⁻ functionally competent β -cell subpopulations all show similar characteristics with incidences of 50-80%. Given our understanding of how heterogeneous cells interact via GJ coupling, it is tempting to speculate that the opposing less functional β -cell subpopulation (i.e., MIP-GFP^{low}, PSA-NCAM^{low}, Fltp⁻, ST8SIA1⁺; 15–50%) could significantly suppress islet activity via electrical coupling. However, Fltp⁻ cells showed a marked reduction in *Gjd2* expression encoding Cx36, which would be expected to reduce their ability to suppress Ca²⁺ elevations across the islet, thus allowing for normal islet function. As such, the increased number of Fltp⁻ cells in development or metabolic stress will likely not impair islet function. Whether MIP-GFP^{low}, PSA-NCAM^{low}, or ST8SIA1⁺ cells show reduced Gjd2/Cx36 expression remains to be determined. Therefore, as a result of functional remodeling, the smaller less functional immature β -cell population within the islet is unlikely to have a detrimental effect on overall islet function. Until recently it has been unclear whether further subdivisions within the functionally competent subpopulation have a disproportionate role within the islet, in part owing to a lack of means to discover these subdivisions.

Optogenetics

Interrogation of β -cell function within islets requires technologies with high spatiotemporal resolution to capture and manipulate signaling events that occur throughout the islet. This can be achieved using optogenetics, where ion channels or pumps derived from light-sensitive bacteria are recombinantly expressed in the cell membrane to optically modulate electrical excitability and β -cell function (38). Combined with high-speed optical imaging, this provides a powerful platform for understanding subpopulation function. Pioneering studies by Reinbothe et al. (38) showed the utility of channelrhodopsin-2 (ChR2), a light-activated cation channel, to modulate islet Ca²⁺ fluxes using blue light.

Later studies used halorhodopsin (eNpHR3.0), an orange light–activated Cl⁻ pump, and ChR2 to selectively silence or activate single β -cells, as well as clusters and larger regions (39,40) (Fig. 3).

Using Optogenetics to Discover and Understand $\beta\mbox{-Cell}$ Subpopulations

Optogenetic mapping has provided evidence that a small subset of functionally competent β -cells may play a disproportionate role in orchestrating islet responses to glucose (Fig. 3) (39). Random activation of single β -cells using ChR2 revealed significant elevation of [Ca²⁺], in neighboring regions in \sim 50% of cases, presumably corresponding to the functionally competent β-cell subpopulation. However, substantial activation was achieved in rare cases (\sim 5%), with Ca²⁺ spreading throughout large portions of the islet. Supporting an important link between glucose metabolism and heterogeneity indicated in prior studies, elevation of Ca²⁺ was most effective in stimulated cells that had the highest NAD(P)H responses (40). Furthermore, a separate subregion of β -cells was identified, characterized by low NAD(P)H responses, that showed high intrinsic oscillatory frequencies and corresponded to initiating Ca^{2+} wave propagation (40). Along similar lines, targeted silencing using eNpHR3.0 highlighted a subpopulation (\sim 1–10%) of metabolically adapted β -cells involved in supporting islet-wide Ca²⁺ dynamics (39). These cells, termed hub cells, were typified by elevated GCK content and hyperpolarized mitochondria, highly consistent with other functionally competent subpopulations. However, they also possessed ~two- to fivefold lower protein expression levels of Pdx1/Nkx-6.1/insulin, indicating relative immaturity, although all three markers were still detectable (Figs. 1 and 3). Whether eNpHR3.0-defined hub cells and ChR2-defined subpopulations are fixed in time or dynamic remains unknown due to the inability to record for more than a few hours, although longitudinal intravital imaging techniques may aid this (41). That these more distinct populations were missed in biomarker analysis or highthroughput single-cell analysis likely indicates the importance of their presence in the intact islet for function and further argues for the importance of studying β -cell heterogeneity in the islet context.

An improvement in metabolic properties and function in hub cells is difficult to reconcile with a loss of β -cell identity. However, it is worth noting that glucose-stimulated insulin secretion in MafA-deficient β -cells (i.e., less mature) is only reduced by ~50% with minimally affected NAD(P)H responses (42), and islets deficient in GLUT2 display preserved second-phase insulin secretion (43). Moreover, recent studies have shown that immature, proliferative β -cells highly express genes involved in amino acid metabolism and mitochondrial function (44). Last, subpopulations of mouse β -cells exist with high *Gck* but low *Pdx1/Nkx-6.1* (20) and low insulin expression (45). Thus, these data indicate a subdivision within the functional population in the islet, raising interesting questions about the role of β -cell identity in islet function and insulin secretion.

Extrapolating Single-Cell Analyses to Islet Studies

Some caution is required when trying to extrapolate results between transcriptional and functional studies. As discussed above, it is well acknowledged that tissue architecture influences gene expression and dissociated cell mRNA profiles may not necessarily reflect those in situ (46). Isolation of single cells is usually associated with decreases in viability. Although these cells are excluded from RNA-Seq analyses of purified populations, some subpopulations, including hub cells (39) or those that express low SERCA2/ low Pdx1 (47), appear more susceptible to endoplasmic reticulum stress and apoptosis/cell death. As such, their gene/ protein expression levels may not be fully captured. Last, there is a tendency to classify and label subpopulations, whereas β -cells likely occupy an overlapping continuum due to the dynamic nature of gene transcription (46). Thus, a major obstacle to our present understanding of heterogeneity is the inability to track transcriptional/ protein dynamics in single β -cells in situ and over long time scales.

Islet Heterogeneity In Vivo

Islets are highly vascularized, and pioneering intravital imaging studies in mice have shown that islet blood flow is directional, perfusing α -cells before β -cells (48). Key to understanding heterogeneity in vivo will be experiments using islets transplanted into the anterior chamber of the eye (41), where blood and neural supplies remain intact. Combined with optogenetics and reporter gene imaging (see below), this provides a powerful means to longitudinally image the plasticity of subpopulations across the life span. However, the anterior chamber of the eye is unable to recapitulate normal pancreatic physiology (e.g., endocrine– exocrine interactions), and immune infiltration has been reported (49).

FUTURE TOOLS FOR THE INTERROGATION OF $\beta\mbox{-}\mbox{CELL}$ HETEROGENEITY

We have argued the importance for resolving and studying β -cell heterogeneity in the context of the intact islet. Although we have discussed several new technologies that have been developed and applied to examine β -cell heterogeneity, additional technologies show great promise for studying β -cell heterogeneity in situ.



Figure 3—Heterogeneity in the intact islet. *A*: Different β -cell subpopulations coexist within the islet, each characterized by different gene/protein expression patterns, morphological markers, glucose responsiveness, insulin secretion, and proliferative capacity and function (Fig. 1). Fltp⁻ proliferative (blue) and Ucn³ transdifferentiating (purple) less functional immature populations are likely nonelectrically coupled and thus do not significantly affect islet-wide responses to glucose. *B*: Optogenetic mapping reveals highly functional β -cell subpopulations with varying identity markers, metabolic properties, and Ca²⁺ responses, but with the ability to exert disproportionate control over coordination and intraislet Ca²⁺ responses. This includes eNpHR3.0-silenced β -cells (hub cells, green) in which halorhodopsin activation and membrane hyperpolarization disproportionally silence the islet and ChR2-activated cells (red) in which ChR2 activation and membrane depolarization disproportionally activate the islet. A population of cells in which ChR2 activation has little effect also shows pacemaker-like characteristics owing to their higher intrinsic oscillation frequency. Adapted from Servier Medical Art under a CC-BY3.0 license (https://creativecommons.org/licenses/by/3.0/).

Photopharmacology

Photopharmacology describes the synthesis of pharmacophores bearing azobenzene photoresponsive elements whose inactive and active states can be controlled by light through *cis-* and *trans-*isomerization. These approaches have been applied to antidiabetes agents including sulfonylureas or incretin mimetics (50). A major advantage of photopharmacology over optogenetics is the ability to control endogenous targets without the need for recombinant approaches, thus providing high utility for targeting human β -cells. However, these approaches tend to be nonbinary, with some drug activity seen even with inactive or unilluminated compound, which can complicate functional interrogation of β -cell signaling.

Matrix-Assisted Laser Desorption/Ionization Imaging

Matrix-assisted laser desorption/ionization (MALDI) is a label-free method that provides information on endogenous proteins, metabolites, lipids, and hormones. This requires application of a matrix to allow absorption of laser energy followed by desorption and liberation of single protonated species, whose time-of-flight (TOF) is ranked according to analyte. Although the spatial resolution of MALDI-TOF is low (20–200 μ m), many endogenous markers can be studied simultaneously in situ with high sensitivity. As such, it may be useful for understanding heterogeneity between individual islets or islet regions (51); however, the technique is temporally limited due to the requirement to use fixed tissue.

Multiplexed Ion Beam Imaging

Mass cytometry overcomes limitations in channel number with conventional flow cytometry by using rare earth–labeled antibodies and TOF detection. This enables large (>40) protein panels to be studied and has been applied to dissociated human β -cells, resolving three distinct subpopulations (16). Technology using rare-earth metal labeling and TOF detection has been adapted into an imaging format multiplexed ion beam imaging (MIBI)—that can achieve subcellular spatial resolution (52). Therefore, imaging of islets in situ, including in intact pancreas preparations, will be achievable. One potential limitation will be the availability of compatible metal-labeled antibodies against proteins involved in β -cell function (e.g., GCK/ion channels/ GPCRs).

Reporter Gene Imaging

Gene expression dynamics (46) may add another level of heterogeneity to β -cell function and potentially explain some of the differences in β -cell subpopulations observed to date. Early studies by Szabat et al. (45) used a Pdx1-mRFP/Ins1-GFP dual reporter to show the existence of a Pdx1⁺/Ins^{low} subpopulation. A similar approach can be envisaged using a combination of fluorophores with long and short half-lives to allow measurement of protein expression dynamics (as an indirect readout for gene expression) in the intact islet setting and to examine plasticity in β -cell subpopulations. However, the disadvantage of such

approaches is the inability to visualize transcription/translation of endogenous genes/RNAs or account for posttranscriptional regulation.

RNA Fluorescent In Situ Hybridization

Fluorescent in situ hybridization (FISH) allows nascent RNA to be labeled, providing a snapshot of transcription. High-throughput single-cell FISH has been applied to examine changes in mRNA levels for β -cell repressed genes during proliferation (53). Closer examination of the FISH data revealed marked differences between individual β -cells (10- to 100-fold), suggesting heterogeneity at the transcript level (53). By using higher resolution single-molecule detection (ffiFISH) to robustly quantify RNA levels in pancreas tissue sections, marked heterogeneity in *Ins2* and *Nkx2–2* was observed between β -cells. This included observing more immature (low *Ins2*, *Nkx2–2*) signatures at the islet periphery (54).

CROP-Seq and Peturb-Seq

Differential gene expression can be obtained in single genome-edited cells residing within a pooled population by directly linking guide RNA expression with individual cell transcriptome responses, for example, via incorporation of polyadenylated sequences recognized by RNA-Seq or combining gRNA with expression of a bar code (55). Both techniques allow high-throughput CRISPR screening of signaling/regulatory pathways in heterogeneous populations and could potentially be used to understand the interactions underlying heterogeneity at different developmental stages.

Computer Modeling

To move beyond a qualitative overview (Fig. 1), computer models show promise in reconciling divergent data sets generated from the range of available technologies ('omics, imaging) and sample preparations (mouse, human, intact, dissociated). Computer models accurately described how a defined β -cell population impacts islet stimulus secretion, e.g., KATP channel mutant (30). Computer models also accurately described stimulus secretion in subpopulations uncovered through ChR2 stimulation (40). We envision that islets could be modeled to incorporate subpopulations of β -cells with characteristics defined by single-cell approaches, and these models could be compared with functional studies. Rapid in silico screening could then identify overlap between subpopulations identified using different experimental approaches and predict how they would influence islet function. A necessary step will be validating theoretical predictions with focused experiments.

SUMMARY

Spurred on by technological advances in single-cell genomics and imaging, the renewed interest in β -cell heterogeneity has changed our view of islet development and function. Pertinently, β -cell subpopulations have been identified that provide deeper mechanistic understanding of poorly characterized processes, such as proliferation,

differentiation/maturation, and stimulus secretion. Thus, β-cell heterogeneity has highlighted new facets of islet plasticity and how changes in this may affect islet function under conditions associated with diabetes. However, most studies to date have investigated dissociated cells, where many transcriptomic/protein features may be altered. Moreover, although the molecular signatures of subpopulations are well characterized, how these influence protein expression and functional state are poorly understood. Key to better understanding heterogeneity will be studies in the intact tissue and even in vivo, the integration of multiple facets of a cell's identity, and models with better fidelity for reading out gene and protein expression dynamics. Indeed, several gaps in our knowledge remain: how do we robustly define a β -cell subpopulation? What is the molecular basis and functional role for heterogeneity among other endocrine cells (α -cells, δ -cells)? Are subpopulations of β -cells (or other cells) fixed into a specific functional state, or are they transitional? Recent and future technological developments should allow these questions to be addressed and, in doing so, may reveal new genes and pathways that can be harnessed to improve therapy, coax β -cell regeneration, and inform de novo engineering of islets and islet transplantation.

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