Limitations in PPARα-dependent mitochondrial programming restrain the differentiation of

human stem cell-derived β cells

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ABSTRACT

Pluripotent stem cell (SC)-derived islets offer hope as a renewable source for β cell replacement

for type 1 diabetes (T1D), yet functional and metabolic immaturity may limit their long-term

therapeutic potential. Here, we show that limitations in mitochondrial transcriptional programming

impede the formation of SC-derived β (SC-β) cells. Utilizing transcriptomic profiling, assessments

of chromatin accessibility, mitochondrial phenotyping, and lipidomics analyses, we observed that

SC-β cells exhibit reduced oxidative and mitochondrial fatty acid metabolism compared to primary

human islets that are related to limitations in key mitochondrial transcriptional networks.

Surprisingly, we found that reductions in glucose-stimulated mitochondrial respiration in SC-islets

were not associated with alterations in mitochondrial mass, structure, or genome integrity. In

contrast, SC-islets show limited expression of targets of PPARα, which regulate mitochondrial

programming, yet whose functions in β cell differentiation are unknown. Importantly, treatment

with WY14643, a potent PPARα agonist, induced expression of mitochondrial targets, improved

insulin secretion, and increased the formation of SC-β cells both *in vitro* and following

transplantation. Thus, PPARα-dependent mitochondrial programming promotes the differentiation

of SC-β cells and may be a promising target to improve β cell replacement efforts for T1D.

INTRODUCTION

All forms of diabetes are associated with reduced mass and function of pancreatic islet β-cells, rendering the body unable to produce sufficient insulin to meet metabolic needs[1]. Thus, replacement of insulin-producing β cells remains a primary goal for diabetes therapy. Pluripotent stem cell (SC)-derived islets, which consist of a heterogeneous mixture of β cells, other islet hormone cell types, as well as immature progenitors and polyhormonal cells, are a promising source for β cell replacement, as these cells could be generated at scale to meet the substantial needs for patients with type 1 diabetes[2,3] (T1D). Directed differentiation of stem cells towards the β cell lineage requires a relatively lengthy (~1 month), complex, multi-stage protocol, including sequential use of small molecules and growth factors, cellular enrichment approaches, and modulation of spatial environments, which support the progression of SCs to progenitor cell types and eventually towards islets[4–8]. Despite the many advances in SC-islet differentiation, these protocols produce a relatively lower frequency of insulin[+] β-like cells that are not fully functional *in vitro* and have limited glucose-stimulated insulin secretion (GSIS) compared to β cells within primary human islets.

Optimal mitochondrial function is essential to maintain GSIS, yet mitochondrial respiration is impaired in SC-islets and is of unclear etiology[5,6,9,10]. Limitations in mitochondrial metabolism of SC-islets also persist following transplantation *in vivo*[6,10,11,12]. Further, several key aspects of mitochondrial metabolism in SC-islets, such as mitochondrial fatty acid metabolism, remain unexplored. Mitochondria are vital to the function of metabolic tissues, yet recent studies position mitochondria as more than client organelles tasked with energy production and have also been shown to regulate cell fate[13,14]. However, whether mitochondrial programming can promote the SC-derived β (SC-β) cell lineage is unknown.

Here, we hypothesized and tested how targeting mitochondrial programming would improve the differentiation and function of SC-β cells. The objectives of our studies were to carefully dissect the mitochondrial metabolic function of SC-β cells, interrogate mitochondrial

programming across β cell differentiation, and explore potential mitochondrial programming

pathways that could be targeted to enhance mitochondrial function and, ultimately, SC-β

differentiation. Together, we speculated our studies would mechanistically reveal how limitations

in mitochondrial metabolism serve as an impediment to optimal SC-β cell differentiation.

RESULTS

SC-islets are transcriptionally and metabolically immature

We generated SC-islets via directed differentiation of the H1 human embryonic stem (hES) cell

line towards a β cell fate utilizing a protocol developed by the Millman laboratory[4] (Fig. 1A and

S1A). To evaluate changes in transcriptional signatures throughout differentiation, as well as

markers of islet maturity, we performed RNA-sequencing (RNA-seq) across all differentiation

stages with comparison to human islet preparations from non-diabetic donors (Fig. S1B-E). These

studies, as well as immunofluorescence and flow cytometry studies, demonstrated that SCs, and

ultimately SC-islets, behaved largely as expected throughout differentiation, with expression of

stage-specific markers and expression of islet-specific hormones in stage 6 (as well as

polyhormonal cells; Fig. 1B-C and S1A). SC-islets also exhibited reduced expression of islet

hormones, with exception of *GHRL*, lower levels of β and α cell maturation markers, relative

hyporesponsiveness of insulin release to glucose- and KCl-stimulation, and lower insulin content

compared to primary human islets (Fig. 1D-F).

To complement our findings of transcriptional and functional immaturity in SC-islets, we

next profiled metabolic endpoints in SC-islets. We first surveyed the glycolytic intermediates

pyruvate and lactate by LC-MS, given the high importance of pyruvate utilization in mature β

cells[15-18]. We found that stage 5 and stage 6 cells increased pyruvate concentrations in response

to glucose stimulation, albeit at lower levels than human islets (Fig. 1G). Moreover, SCs at every

stage generated significantly higher quantities of lactate, including in stage 6 cells, as compared

to human islets (Fig. 1H). The presence of higher quantities of lactate in SC-islets than human

islets could be consistent with previous observations of defects in glycolytic flux, an inability to

silence lactate metabolism due to expression of disallowed genes (such as *LDHA*), or due to the

inability to switch from primarily glycolytic to oxidative metabolism in immature β cells, all of which

have been reported previously[5,6,10,18,19]. We next performed respirometry studies utilizing the

recently described Barofuse instrument, which allows for the assessment of respiration in a gas

pressure flow-based perifusion style platform[20]. We confirmed that SC-islets did not increase

oxygen consumption upon glucose stimulation as compared to human islets, which has been

shown previously[5,6,10] (Fig. 1I). We also observed that SC-islets displayed more sensitivity to the

effects of the ATP synthase inhibitor oligomycin, which is expected to lower oxygen consumption,

and the uncoupler FCCP, which transiently increases oxygen consumption, than primary human

islets, possibly suggesting that SC-islets may not be as uncoupled as primary human islets but

could also be related to relative differences in penetration of the compounds (Fig. 1I). Taken

together, SC-islets display a signature of transcriptional, functional, and metabolic immaturity.

113   Mitochondrial DNA content, mass, and architecture in SC-islets are comparable to those

114   of human islets.

Optimal β cell function is highly dependent on mitochondrial quality and health[21–23]. Given our

observations of transcriptional, functional, and metabolic immaturity in SC-islets, we profiled

mitochondrial mass and structure of SC-islets at different stages of differentiation and human

islets to obtain insights into limitations in mitochondrial respiration as recent studies have had

discordant results regarding reductions in mitochondrial mass in SC-islets compared to primary

human islets[5,10]. To survey mitochondrial genome integrity, we measured mitochondrial (mt)DNA

content, and while we found higher mtDNA content in stage 0 embryonic stem cells, we noted no

significant differences between stage 1-6 cells and human islets (Fig. 2A). To next determine if

there were deficiencies in mtDNA replication, we evaluated the initiation of first strand replication.

In mtDNA replication, first strand synthesis begins with formation of the D-loop. Elongation of the

7S sequence, which is contained in the D-loop, into the nearby region, *Cytb*, signifies committed

initiation of first strand replication. We assessed both D-loop formation and first strand replication

initiation and found no differences between stage 0-6 cells and human islets (Fig. 2B). We also

observed no differences in mitochondrial mass in stage 0-6 cells compared to human islets by

use of two complementary approaches; namely, citrate synthase activity and TOM20 protein

expression (Fig. 2C-D and S1F). To assess mitochondrial ultrastructure, we next performed

transmission electron microscopy. While we noted a lower frequency of mature insulin granules,

as previously described[5,6,24,25], we did not observe any differences in quantitative measures of

mitochondrial architecture in stage 6 SC-β cells compared to primary human β cells (Fig. 2E).

Together, these results suggest that limitations in mitochondrial respiration in SC-islets were not

a consequence of reduced mitochondrial genome integrity, DNA replication, mass, or structure.

Transcriptomic profiling reveals potential limitations in mitochondrial metabolism in SC-

islets

Due to the similarities in mitochondrial genome integrity, DNA replication, mass, and structure

between SC-islets and human islets, we next explored potential transcriptional differences that

might provide mechanistic insights into limitations in glucose-stimulated mitochondrial respiration

in SC-islets. Following a comparison of RNA-seq data between stage 6 SC-islets and human

islets, we found that 11,858 genes were significantly differentially expressed, including 405 genes

represented within MitoCarta 3.0[26], a curated set of nuclear and mitochondrial-encoded genes

with high confidence of localization to the mitochondria (Fig. 3A). While we observed dynamic

changes in expression of MitoCarta genes throughout differentiation, we noted a subset of highly

expressed MitoCarta genes in primary human islets that appeared to remain relatively lowly

expressed throughout SC differentiation, including by stage 6 (Fig. 3B). Pathways analysis of

differentially expressed MitoCarta genes showed differences in fatty acid metabolism, amino acid

metabolism, and oxidative phosphorylation (OXPHOS) (Fig. 3C-D). As there were a greater

number of downregulated Mitocarta 3.0 genes in SC-islets compared to human islets, we wanted

to understand the possible biological processes associated with these gene targets via the Search

Tool for Recurring Instances of Neighboring Genes (STRING) database[27]. Following STRING

analysis of significantly downregulated MitoCarta genes from stage 6 SC-islets, gene ontology

(GO) enrichment analysis was used to clarify pathways related to connections identified by

STRING. Pathways identified were related to mitochondrial translation, the electron transport

chain (ETC)-OXPHOS system, and lipid metabolism (Fig. 3E). Collectively, our transcriptomic

studies revealed several pathways that could contribute to limitations in mitochondrial metabolism

in SC-islets.

Limitations in the OXPHOS machinery and mitochondrial lipid metabolism in SC-islets

We next wished to clarify the specific nature of transcriptional defects affecting mitochondria in

SC-islets. An initial question was whether the limitations in SC-islet mitochondria were related to

an impairment in the shift of the glycolytic to OXPHOS metabolic program, which was previously

observed in human induced pluripotent stem cell (iPSC)-derived islets[11]. While we observed that

expression of genes related to glucose transport and glycolysis were enriched in stage 0

pluripotent cells, we did not observe differences in the glycolytic gene expression program

between stage 6 SC-islets and human islets (Fig. 4A). We also did not observe a significant

difference in the overall RNA expression of genes encoding the entirety of TCA cycle enzymes or

enzymes/transporters involved in lactate metabolism between stage 6 SC-islets and human islets,

whereas these genes were more highly enriched in stage 0 ES cells (Fig. S1G). We next

considered the possibility that mitochondrial impairments in SC-islets could be related to

limitations in the β cell OXPHOS machinery. Indeed, we observed a notable subset of 29

OXPHOS gene targets that were enriched in human islets compared to stage 6 SC-islets (Fig.

4B-C). These upregulated targets in human islets included genes encoding subunits of all 5 ETC

complexes, as well as complex assembly factors (Fig. 4B-C). Notably, stage 0 pluripotent cells

also had an enrichment of a subset of OXPHOS gene targets compared to stage 1-6 cells;

however, these genes appeared to be largely distinct from those upregulated within human islets

(Fig. 4B). To confirm that this phenotype was specific to SC-β cells compared to other SC-islet

cell types, we re-analyzed data from a recently published single-cell RNA-seq dataset comparing

SC-islets and human islets, as it has been observed that the various directed differentiation

protocols from independent laboratories produce SC-islets with similar transcriptomic profiles[28,29].

Indeed, we observed that >50% of OXPHOS gene targets were lower in SC-β cells compared to

primary human β cells, whereas there was no significant differences in glycolytic gene expression

between the groups (Fig. 4D). Reduced expression of OXPHOS genes appeared to be β cell-

specific, as we did not find robust differences in SC-α cells compared to primary human α cells

(Fig. 4D). Lower expression of ETC and OXPHOS gene targets in SC-β cells was also supported

by transcriptomic profiling by others[6]. We next measured expression of subunits of OXPHOS

complexes via Western blot and found that their expression was significantly lower in SCs

compared to human islets across all differentiation stages (Fig. 4E). Thus, we observed lower

expression of components of the OXPHOS machinery in SCs compared to human islets that

appear to be distinct from programming of the glycolytic gene network.

While β cells are not classically considered central regulators of physiologic lipid storage

and turnover, lipid metabolism and signaling are well-known to be essential for the maintenance

of optimal β cell insulin release[30,31]. Importantly, mitochondrial lipid metabolism has not been

examined in SC-islets to date. We observed distinct subsets of genes related to fatty acid

biosynthesis and β-oxidation that were enriched in human islets but were not increased during

differentiation towards stage 6 SC-islets (Fig. 5A-B). Specifically, we found genes encoding key

initial steps in fatty acid biosynthesis and β-oxidation were markedly lower in SC-islets compared

to human islets, including *ACACB*[32,33], a key component of acetyl CoA carboxylase, which

catalyzes conversion of acetyl CoA to malonyl CoA as an early step in fatty acid biosynthesis,

CPT1A[32,34], the rate-limiting enzyme of fatty acid oxidation that catalyzes the transfer of the long-

chain acyl-CoA ester to carnitine, to allow fatty acids to enter the mitochondrial matrix for

oxidation, and *ACAT1*[34,35], which catalyzes the final step of mitochondrial β-oxidation (Fig. 5C).

To assess if differentiating SCs and SC-islets possessed alterations in mitochondrial fatty

acid metabolism, we examined acylcarnitine species by LC-MS, as carnitine is a required cofactor

for lipid transport into the mitochondria for fatty acid β-oxidation[36]. Notably, we could not detect

long-chain acylcarnitines or L-carnitine during differentiation, while these were detectable in

human islets (Fig. 5D-E). We next noted low levels of short-chain acylcarnitines throughout SC

differentiation, with elevations in short-chain acylcarnitines by stage 6. However, both short- and

long-chain acylcarnitines were significantly reduced throughout differentiation to SC-islets when

compared to human islets (Fig. 5E). Furthermore, we observed lower expression of *ALDH9A1*, a

gene encoding a carnitine biosynthetic enzyme, as well as in *CPT1A*, the key regulator of long-

chain acylcarnitine synthesis, in SC-islets compared to human islets, yet other regulators of short-

chain acylcarnitine synthesis and other genes in the carnitine biosynthetic pathway were not lower

in SC-islets (Fig. 5C and 5F). Of note, L-carnitine is not included within either SC-islet[4] or human

islet media formulations, thus these differences were unrelated to L-carnitine supplementation (B.

Haight, Prodo Laboratories, personal communication as human islet media formulation is

proprietary). These data indicate that in addition to limitations in expression of the ETC-OXPHOS

system, SC-islets exhibit robust differences in mitochondrial fatty acid metabolism compared to

primary human islets.

To ensure our transcriptomic findings were not limited to differentiation of a single hES

line, we performed RNA-seq in stage 0-6 cells generated from the HUES8 hES cell line and

compared these results to primary human islets. We again observed the expected expression of

stage-specific differentiation markers and limitations of islet hormones and of β and α cell

maturation markers in stage 6 cells, similar to our observations in H1 cells (Fig. S2A-C). Further,

we observed lower expression of key regulators related to fatty acid β-oxidation, glucose transport

and glycolysis genes, and genes encoding subunits of all 5 ETC complexes, complex assembly

factors in stage 6 SC-islets compared to primary human islets (Fig. S2D-F). In sum, these data

indicate that SC-islets have several limitations in mitochondrial programming, including within the

ETC-OXPHOS system and mitochondrial fatty acid metabolism.

SC-islets display lower expression of PPARα and PPARγ transcriptional targets

Due to the transcriptional differences in the OXPHOS machinery and mitochondrial lipid

metabolism, we hypothesized that limitations in key transcriptional pathways might prevent

optimal mitochondrial programming in SC-islets. To identify candidate transcriptional regulators,

we employed the bioinformatic algorithm Enrichr[37–39] to analyze differentially expressed Mitocarta

3.0 genes between stage 6 SC-islets and human islets (Fig. 6A-B). This analysis yielded several

expected findings, including the estrogen-related receptor (ERR) family, PDX1, and PRDM16,

which all have been associated with the regulation of β-cell development, as well as several

effects on mitochondria, including mitochondrial biogenesis, mitophagy, or induction of oxidative

metabolism[11,40–43]. We were also intrigued by the enrichment of PPARα and PPARγ in SC-islets,

nuclear receptors that are both tightly connected with mitochondrial function, including the

regulation of fatty acid oxidation, mitochondrial biogenesis, and coordination of mitochondrial

programming together with the co-activator PGC1α, a central regulator of mitochondrial

metabolism[44–46]. Thus, we next assessed the expression of PPARα and PPARγ target genes in

stage 6 SC-islets and human islets as PPARα and PPARγ had not been previously studied during

SC-β cell differentiation. We found that expression of canonical PPARα and PPARγ target genes

were lower in stage 6 SC-islets compared to human islets (Fig. 6C). Further, PPARα regulates

carnitine biosynthesis and transcriptionally regulates expression of *ALDH9A1*[47,48], which is also

lower in SC-islets compared to human islets, as noted above (Fig. 5F).

We next wished to determine the mechanisms by which PPAR transcriptional pathways

are lower in SC-islets. PPARα classically complexes with RXR, and together with PPAR and RXR

ligands and transcriptional co-activators, form a transcriptional complex that can bind PPAR

response elements to facilitate target gene expression[49]. Thus, we first assessed expression of

the PPAR and RXR transcriptional machinery in stage 6 SC-islets. We found that *PPARA*,

PPARG, *RXRA*, and *RXRB* expression levels were higher in stage 6 SC-islets compared to

human islets (Fig. 6D). While we found that PPARα protein levels appeared to rise throughout

differentiation, PPARα protein levels appeared unchanged in stage 6 SC-islets compared to

human islets (Fig. 6E). The expression of the PPAR-related co-activators *NCOA1*, *EP300*, *EIF4E*,

SRC, and *CTNND1* were also more highly expressed in stage 6 SC-islets compared to the human

islets (Fig. 6D). To determine if these findings also occurred specifically in β cells, we reanalyzed

a publicly-available single-cell RNA-seq data set[28] and found that expression of the *PPARA* and

RXRA were significantly higher in SC-β cells compared to primary human β cells (Fig. 6F).

Further, while we observed lower expression of (*PPARG1A*/*PGC1α*) in SC-islets, we did not

observe differences in *PGC1α* expression between SC-β cells and primary human β cells (Fig.

6D and F). To assess if chromatin accessibility at putative PPAR- and RXR-binding sites was

affected in SC-β cells, we next reanalyzed publicly-available single-cell ATAC-seq data from SC-

β cells and primary human β cells[28]. We found that PPAR and RXR motifs were highly accessible

in SC-β cells compared to primary β cells (Fig. 6G). These results suggest that limitations in the

PPARα/γ transcriptional program in SC-β cells was not due to a loss of the PPAR transcriptional

machinery or chromatin accessibility.

We next identified a differentiation trajectory of stage 6 SC-β cells using publicly-available

single-cell RNA-seq and single-cell ATAC-seq data[28] to understand the dynamics of PPARα

expression and activity during β cell differentiation. Using single-cell RNA-seq expression data

specifically from stage 6 SC-islet cells, we first identified a developmental trajectory from

immature *INS*+ β cells co-expressing the key endocrine progenitor transcription factor *FEV* to more

mature β cells with higher *INS* expression and low/absent expression of *FEV* (Fig. S3A-B). Stage

6 SC-β cell trajectory was then extracted and combined with single-cell RNA-seq data from

primary human β cells to compare expression. We then evaluated several β cell transcriptional

regulators, observing greater expression in *PDX1* and *PAX6* across the SC-β cell differentiation

trajectory (Fig. S4A). Similarly, *PPARα* expression increased along the SC-β cell differentiation

trajectory (Fig. 6H). Despite this, we observed that the PPARα gene signature score, calculated

using the expression of 51 known PPARα transcriptional targets[50], remained at a lower level than

primary human β cells throughout the SC-β cell differentiation trajectory (Fig. 6I). We next used

ChromVAR[51], a computational approach to estimate a per-cell bias-corrected measure of the

change in chromatin accessibility in sets of peaks containing the same transcription factor motif

compared to an average cell profile, which serves as a proxy for motif activity. Again, an evaluation

of known β cell transcriptional activators revealed increases in *NEUROD1* and *PAX6* motif

deviation across SC-β cell differentiation trajectory (Fig. S4B). Intriguingly, we observed that the

estimated PPARα motif deviation did not change significantly across the SC-β cell differentiation

trajectory but was consistently greater than that of primary β cells (Fig. 6J). Taken together, our

analysis of bulk RNA-seq, pseudo-bulk single-cell RNA-seq and single-cell ATAC-seq, and

transcriptomic and chromatin accessibility profiling across the SC-β cell differentitation trajectory,

all revealed that SC-β cells possess lower expression of target genes within the PPARα pathway

that do not appear to be due to reduced expression of the PPAR transcriptional machinery,

accessibility of PPAR-binding motifs, or PPARα motif activity.

WY14643 improves the differentiation of SC-β cells

While the actions of PPARα/γ are well known in many cell types, including studies in postnatal β

cells[52,53], the importance of PPARα/γ during β cell differentiation are unclear. We speculated that

the limited activation of PPARα/γ-related mitochondrial programming could be related to the lack

of PPARα/γ ligands during SC-islet differentiation. Notably, the primary ligand for RXR is retinoic

acid, which is routinely included in SC-islet differentiation protocols[4,54]. Ligand binding to PPARs

induces a conformational change that displaces co-repressors and enhances binding to co-

activators to promote the expression of target genes[55,56]. PPAR agonists have also been shown

to not be required to increase the number of PPARγ-binding sites, suggesting that chromatin

accessibility at PPAR motifs in SC-β cells could be maintained and poised for activation in the

absence of an agonist[57], consistent with our observations above (Fig. 6F and I). Therefore, we

hypothesized that addition of a PPAR agonist may improve SC-β cell differentiation.

As a proof of concept, we chose to include the potent PPARα agonist WY14643[58–61] during

stage 6 of the differentiation program given our observations of a relative lack of PPARα target

gene expression throughout the SC-β cell maturation trajectory despite a maintenance of PPARα

motif activity (Fig. 6I-J and 7A). Intriguingly, we found that WY14643 treatment significantly

increased the percentage of C-peptide-positive SC-islets without changes to the glucagon-,

somatostatin-, or polyhormone-positive cell populations (Figs. 7B). WY14643 treatment also

significantly raised the frequency of NKX6.1 and C-peptide co-positive SC-β cells with similar

increases in PDX1 and C-peptide co-positivity (Fig. 7C-D).

To gain further insight into the mechanisms underlying the expansion of SC-β cells upon

PPAR agonism, we performed bulk RNA-seq studies on WY14643-treated SC-islets compared to

control-treated SC-islets. Firstly, we observed the expected induction of a subset of PPARα

targets in WY14643-treated SC-islets compared to control-treated SC-islets (Fig. 7E). We next

observed higher expression of genes reported to be expressed in mature β cells[62] upon WY14643

exposure, including *INS*, *UCN3*, *GCK*, and *G6PC2*, as well as lower expression of genes enriched

in immature β cells[62], which could also be reflective of a higher percentage of β cells overall in

bulk assessments of WY14643-treated SC-islets (Fig. 7F-G). In addition, we found greater

expression of ETC-OXPHOS-related genes, with minimal changes in expression of genes related

to glucose uptake and glycolysis, in WY14643-treated SC-islets compared to control-treated SC-

islets (Fig. 7H-I). Further, we found that addition of WY14643 significantly improved insulin

secretion in SC-islets compared to controls, yet the fold induction of insulin secretion by glucose

was similar between the groups (Fig. 7J). Together, these studies suggested that WY14643

improves the formation of SC-β cells.

We next evaluated potential mitochondrial mechanisms promoting SC-β cell formation

following WY14643 exposure. We first found modest increases in mitochondrial respiration

following high glucose stimulation in WY14643-treated SC-islets (Fig. 7K). We observed that

WY14643 treatment resulted in greater protein expression of the complex I subunit NDUFA9

compared to controls, however, changes in other OXPHOS subunits were not found (Fig. 7L and

S5A). We also observed no changes in mitochondrial mass, expression of regulators of

mitochondrial dynamics, or mitochondrial ultrastructure following WY14643 exposure (Figs. 7L

and S5B-D). We next found that that WY14643 exposure reduced both fatty acid and acylcartinine

concentrations in SC-islets, consistent with the induction of fatty acid oxidation following PPARα

agonism (Fig. S6A-B). To evaluate the contributions of mitochondrial pyruvate transport or long

chain fatty acid oxidation to the effects of WY14643 on SC-β cell formation, we exposed stage 6

SC-islets to WY14643 (or vehicle control) for 7 days in the presence or absence of the

mitochondrial pyruvate carrier (MPC) inhibitor UK5099 or the CPT1A inhibitor etomoxir. Indeed,

UK5099 exposure abrogated the effects of WY14643 to increase the frequency of SC-β cells,

while etomoxir exposure increased the presence of GCG+ SC-α cells (Fig S6E-F). Finally, we

examined other metabolic signaling pathways, including Calcium/calmodulin-dependent protein

kinase II (CaMKII), AMP-activated protein kinase (AMPK), and acetyl co-A carboxylase (ACC),

yet did not observe differences in the phosphorylation and activation of these pathways following

WY14643 exposure (Figure S6G-I). Taken together, these results suggest that WY14643 induces

both mitochondrial oxidative and fatty acid metabolism to promote SC-β cell formation.

We next wanted to determine if the beneficial effects of WY14643 were visible in other SC

lines and also durable *in vivo* following transplantation. We first tested the effects of WY14643 in

stage 6 SC-islets differentiated from the 19-9-11 induced pluripotent stem cell (iPSC) line (Figure

S7A-B). We observed that addition of WY14643 for 7 days increased the insulin secretion index

of iPSC-derived SC-islets (Fig. S7C). Next, we treated SC-islets (from the H1 ESC line) with

WY14643 or vehicle control for 7 days *in vitro* prior to seeding SC-islets on a poly(lactide-*co*-

glycolide) (PLG) microporous scaffold prior to transplantation into the epididymal fat pad of

immunodeficient NOD-*scid* IL2Rg[null] (NSG) mice. We delivered SC-islet grafts to the epididymal

fat pad rather than the renal subcapsular space as the epididymal fat pad is a translationally

relevant transplant site[63,64] (Fig. 8A). We then retrieved the scaffolds 10 days posttransplant due

to the limited half-life of WY14643, and we did not administer WY14643 *in vivo* to avoid systemic

effects of the agonist. Following graft retrieval, we observed SC-islets appeared as expected

within the scaffold with recipient cell growth within and surrounding the xenograft (Fig. 8B).

Notably, we observed a durable effect of WY14643 on the SC-β cell lineage 10 days post-

transplantation, with a significant increase in C-peptide-positive SC-β cells compared with vehicle-

treated controls (Fig. 8C-D). Further, we observed that WY14643 did not affect the percentage of

somatostatin- or glucagon-positive cells but did lead to a modest reduction in the immature

polyhormone-positive cell population post-transplantation (Fig. 8C-D). Moreover, WY14643

treatment led to a significant increase in insulin content within the explanted scaffold graft

compared to vehicle-treated controls (Fig. 8E). We also observed that WY14643 treatment did

not significantly alter the frequency of Ki-67/C-peptide co-positive cells, suggesting WY14643

likely did not enhance SC-β cell replication (Fig. 8F). WY14643 treatment also did not alter

mitochondrial morphology in SC-β cells post-transplantation (Fig. 8G). Taken together, the

addition of a PPARα agonist to stage 6 SC-islets improved the formation of SC-β cells, which was

durable *in vivo* post-transplantation.

DISCUSSION

Here, we reveal a vital importance for mitochondrial programming to promote the differentiation

of SC-β cells. Our studies elucidate specific mechanisms underlying limitations in SC-β cell

mitochondrial metabolism, including reductions in expression and function of the ETC-OXPHOS

system and mitochondrial fatty acid metabolism. We observe limitations in glucose-stimulated

respiration in SC-islets without alterations in mitochondrial mass, structure, mtDNA content, and

mtDNA replication. We identify several candidates that may contribute to limitations in

mitochondrial programming during SC-β cell maturation, notably the nuclear receptors PPARα

and PPARγ. We demonstrate that the PPARα and PPARγ transcriptional machinery is intact and

chromatin at the sites of PPARα and PPARγ binding motifs is accessible, but expression of PPARα

and PPARγ target genes is reduced, suggesting a deficiency of PPAR ligands in the SC-islet

differentiation process. To test a proof of concept of the importance of PPAR-mediated

programming in SC-β cells, we found that the PPARα agonist WY14643 improved mitochondrial

respiration, fatty acid oxidation, and importantly, insulin secretion. Moreover, WY14643 increased

the formation of C-peptide-positive SC-β cells both *in vitro* and upon transplantation within a

translationally relevant engraftment site, as well as enhancing insulin content within SC-islet

grafts. Thus, our data reveal the potential for targeting PPARα-dependent mitochondrial

programming to improve SC-β cell differentiation, expanding beyond classical roles for

mitochondria simply as a client organelle necessary for β cell bioenergetics.

Our results suggest that the activation of mitochondrial transcriptional programs can

expand the SC-β cell lineage. While the synchronistic connections between mitochondria and cell

fate are well appreciated in other tissue types, such as brown adipocytes[65,66], this connection has

not been clear during β cell differentiation. Interestingly, PPARα/γ regulate the differentiation of

adipocytes, keratinocytes, bone, and several immune cell types and promote the expression of

genes related to lipid metabolism and mitochondrial biogenesis[67–73], yet their function in SC-β cell

differentiation had not previously been observed. In adult human islets, PPARα overexpression

or treatment with a PPARα agonist offered protection from lipotoxicity and preserved GSIS[74,75].

Similarly, PPARγ agonists preserved β cell function in humans with T2D and protected human

islets from lipotoxicity[52,76–78]. However, a role for mitochondrial programming by PPARα/γ to

promote human β cell differentiation was not examined. Our studies reveal the potential for

targeting PPARα to improve mitochondrial respiration and fatty acid metabolism and expand the

SC-β cell lineage. Indeed, we speculate that PPARα could act through mitochondria to promote

β cell differentiation by supporting the bioenergetic demands of differentiation or through

increased substrate availability to facilitate histone modifications and chromatin accessibility at

key β cell differentiation genes, which remains to be tested in the future. Future work will also be

needed explore the addition of PPAR agonists in other SC-β cell differentiation stages, define the

roles of PPARs on metabolism throughout β cell development, and to explore the potential for

shared PPAR-related mechanisms underlying differentiation across tissues, including SC-β cells

as well as adipocytes, keratinocytes, bone, and immune cells.

PPARα/γ are well known regulators of lipid metabolism, and PPARα is a central regulator

of fatty acid oxidation. Our studies suggest that WY14643 enhances both mitochondrial fatty acid

oxidation and respiration in SC-islets. Beyond their roles in lipid metabolism and mitochondrial

biogenesis, emerging studies demonstrate crosstalk between ERRα and PPARα, which could

suggest the potential that WY14643 might indirectly support expression of OXPHOS targets

through ERRs[44,79]. Our studies define a previously unknown impairment in mitochondrial lipid

metabolism in SC-islets with beneficial roles of WY14643 on SC-β cell formation, as well as

OXPHOS gene expression, which appears distinct from the roles of ERRγ to improve SC-β cell

GSIS through induction of OXPHOS and suppression of glycolysis[11]. Future studies will be crucial

to discriminate both the specific and integrated contribution of each of these highly integrated

mitochondrial transcriptional regulators on SC-islet development, including PPARα, PPARγ,

ERRγ, and PGC1α. The impact of pharmacologic agonists of ERRγ[80] and specific, partial, or dual

agonists of PPARα and PPARγ[81–83], to together improve both SC-β cell differentiation (through

PPARα) and maturation/GSIS (through ERRγ) will be intriguing to assess for future translational

benefit.

Our studies expand upon the metabolic differences observed in SC-islets compared to

primary human islets, which have understandably focused to date on glucose metabolism given

the pivotal importance of glucose metabolism for β cell insulin secretion. We found that SC-islets

do not increase mitochondrial respiration upon high glucose stimulation, similar to the

observations of others[6,10]. We noted that SC-islets exhibited both lower oxygen consumption

following oligomycin exposure and a higher peak oxygen consumption after the uncoupler FCCP

compared to human islets, which had not been observed previously. These effects could be a

function of differences of tissue penetration of pharmacologic agents between SC-islets and

human islets despite their similar size and appearance. Alternatively, the differences in SC-islet

responses to oligomycin and FCCP could be indicative that SC-islets may be more coupled or

possibly have reductions in proton leak compared to primary human islets. Primary β cells are

known to be remarkably uncoupled, and blockade of adenosine nucleotide translocase (ANT)-

mediated basal proton leak impairs GSIS[84–87]. Indeed, SC-islets express significantly lower levels

of ANT than primary human islets (log_2FC -0.66, FDR < 0.05). We also observe alterations in

mitochondrial acylcarnitine profiles in SC-islets, indicative of limitations in mitochondrial lipid

metabolism that may originate due to limitations in mitochondrial programming by PPARα/γ. A

recent study demonstrated that elevations in the sphingolipid ceramide limited the maturation of

SC-β cells, suggesting that other lipid subtypes may play distinct roles in differentiation[88].

Reduced levels of C3 and C5 acylcarnitines may also reflect alterations in branched chain amino

acid metabolism[89], an effect we also observed in SC-islets (Fig. 3D). Notably, fluctuations in amino

acid availability at birth may also impact the nutrient sensitivity of mTORC1 signaling pathway to

affect β cell maturation[90]. Unsaturated fatty acids can serve as endogenous ligands for PPARα

and/or PPARγ, leading us to speculate a possible similar role for these metabolites in β cell

maturation, especially given the fatty acid composition of maternal breast milk[91–94]. It will be

intriguing to test if future approaches to modulate β cell uncoupling or nutrient exposure will have

a beneficial impact on SC-β cell differentiation and maturation. On the other hand, exploration of

the SC-islet microenvironment or culture conditions, could provide an alternative approach to

clarify the metabolic control of SC-β cell differentiation and function.

461   Limitations in our study included use of RNA-seq data performed on bulk SC-islet

differentiations, which have a different cell type composition than bulk primary human islets[28]. Our

conclusions were supported, however, by re-analysis and inclusion of published single cell RNA-

seq of SC-β cells and primary β cells[28]. Further, despite reports of mitochondrial dysfunction in

SC-islets from numerous groups, the effect of PPAR agonists will need to be evaluated in other

SC-β cell differentiation protocols, as our observations could be protocol-dependent. While

numerous studies, including genetic knockout models, indicate a specific role for WY14643 both

in vitro and *in vivo* through PPARα[59–61], WY14643 has been reported to have weak PPARγ activity

in vitro[95]. Thus, additional studies to dissect the contribution of PPARγ on SC-islet differentiation

would be valuable in the future. An additional limitation of our study is the need to determine if

PPAR agonism, such as by WY14643 or other agents, is able to promote SC-β cell differentiation

and function to ultimately improve glycemic control upon engraftment in diabetic mice *in vivo*. We

also did not explore the role for systemic administration of PPAR agonists post-transplantation,

which might provide additional opportunities to enhance *in vivo* SC-β cell function.

475   Our findings support metabolic limitations previously found in the generation of SC-β cells,

but also open new roles for of PPARα and mitochondria in the regulation of SC-β cell

differentiation. Our results reinforce the findings of several studies of SC-β cell metabolism,

including defects in OXPHOS expression, persistence of aberrant glycolytic metabolism, and

limitations in expression of the ERR transcriptional pathway[6,10,11], while clarifying new roles for

mitochondrial fatty acid metabolism and OXPHOS mediated at least in part by PPARα. Our results

also demonstrate the novel potential to activate PPARα as a highly druggable target to improve

SC-β cell differentiation. Taken together, our results corroborate limitations that remain in the

formation of SC-β cells and identify new potential routes to overcome these deficits, possibly

requiring pharmacologic targeting of several complementary pathways to generate higher

quantities of fully functional SC-β cells for enhanced therapeutic use in diabetes. Early clinical

trials have provided excitement regarding the therapeutic potential for SC-islet transplantation to

provide a functional cure for T1D[3,96]. However, these allografts require up to 6 months to optimize

their function, and the long-term viability of these cells remain unclear[3]. Thus, our studies highlight

the opportunity to focus on mitochondrial programming pathways to improve the differentiation,

metabolism, and function of SC-β cells to more closely resemble primary β cells as an approach

to improve durable cell-based therapies for T1D.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Stem cell-derived islets (SC-islets): SC-islets were differentiated from H1, HUES8, and IPS-DF 19-9-11 T.H (feeder independent; WiCell) human pluripotent stem cells (hPSCs) using a 6-stage protocol, as reported previously[4,63,64,97]. Briefly, undifferentiated cells were dispersed into single cells, and 5×10^6 cells were seeded in 6-well cell culture treated plates coated with Matrigel (Corning). Cells were cultured for 24 h in mTeSR1 and then cultured in the differentiation media for 6 stages. During each stage, cells received daily media changes with stage-specific basal media supplemented with growth factors and small molecules to induce differentiation. The stages of this protocol are defined as definitive endoderm (stage 1), primitive gut tube (stage 2), pancreatic progenitor (stages 3 and 4), endocrine progenitor (stage 5) and immature stem cell derived β-cell (stage 6). At the end of each stage, cells could be dissociated from the plate, fixed and analyzed using intracellular flow cytometry for stage-specific biomarkers. For reaggregation from planar culture, cells were dispersed into single cells using TrypLE and seeded onto suspension culture 6-well plates at 8×10^6 cells per well. Y27632 was included in the media for the first 24 hours of culture, then media was changed every other day for the duration of culture. Plates were placed on an orbital shaker set to 100 rpm inside a humidified incubator. Reaggregted cell clusters were treated with 60 µM WY14643 alone or with 4 µM UK5099 or 25 µM Etomoxir days 3-10. For scaffolds, reaggregated cell clusters were collected and seeded on the scaffold in the same manner as described above after 3–4 days in culture,.

Microporous scaffold fabrication: Poly(lactide-*co*-glycolide) (PLG) microporous scaffolds were fabricated as previously described[63]. Briefly, PLG microporous scaffolds were fabricated by compression molding PLG microspheres (75 : 25 mole ratio d,l-lactide to glycolide) and 250 to 425 µm salt crystals in a 1:30 ratio of PLG microspheres to salt. The mixture was humidified in an incubator for 7 minutes. Scaffolds were compression molded with 77.5 mg of polymer–salt mixture into cylinders 5 mm in diameter by 2 mm in height using a 5 mm KBr die (International Crystal Laboratories, Garfield, NJ) at 1500 psi for 30 seconds. Molded constructs were gas foamed in

800 psi carbon dioxide for 16 hours in a pressure vessel. The vessel was depressurized at a

controlled rate for 30 min. On the day of cell seeding, scaffolds were leached in water for 1.5

hours, changing the water once after 1 hour. Scaffolds were disinfected by submersion in 70%

ethanol for 30 seconds and rinsed multiple times with phosphate-buffered saline (PBS).

Scaffold seeding of SC-islets: To initiate scaffold culture differentiations from reaggregated

clusters, reaggregated clusters were seeded on scaffolds at a density of 125×10^6 cells/cm^3. Prior

to seeding, scaffolds were washed in cell media solution then briefly dried on sterile gauze to

improve the absorption of the cell solution into the scaffold. Cells were distributed across both

faces of the scaffold and then incubated for 2 hours to allow cell solution to be further absorbed

into the scaffold before differentiation media was added.

Human islets: All human samples were procured from de-identified donors without diabetes from

the Integrated Islet Distribution Program, Alberta IsletCore, or Prodo Laboratories and approved

by the University of Michigan Institutional Review Board. Human primary islets were cultured at

37°C with 5% CO_2 in PIM(S) media (Prodo Laboratories, Aliso Viejo, CA, USA) supplemented

with 10% FBS, 100 U/mL penicillin/streptomycin, 100 U/mL antibiotic/antimycotic, and 1 mM

PIM(G) (Prodo Laboratories, Aliso Viejo, CA, USA). Islets were used from male and female

donors, and donor information is provided in Table S1.

Transplantation of reaggregated SC-islets in immunodeficient mice: SC-islets were

transplanted as previously described[63]. All mice were maintained in accordance with the

University of Michigan's Institutional Animal Care and Use Committee under specific pathogen-

free conditions and were assigned randomly to different experimental groups. Up to 5 mice were

housed per cage and were maintained on regular chow with *ad libitum* access to food and water

on a 12 h light-dark cycle. 8–10-week-old male NOD.Cg-Prkdc[scid] Il2rgtm1Wjl/SzJ (NSG) mice

(Jackson Laboratories) were used for transplantation studies. On the day of transplantation, the

abdomen was shaved and sterilized. A small incision was made in the peritoneal wall and the

epididymal fat pads were unwrapped outside the body. As described above, reaggregated cell

clusters were collected and seeded on both faces of the scaffold and incubated at least 2 h before

transplantation. Reaggregated cell-laden scaffolds were placed on the fat pads, wrapped in fat

pad tissue, and placed back into the peritoneal cavity. Each mouse received approximately 5

million cells in clusters on each scaffold, equaling a total of 10 million cells transplanted per

mouse. Mice were sutured, surgically stapled, and received carprofen (0.5 mg/mL) the day of and

after surgery. Body weight was measured daily for 10 days prior to euthanasia and subsequent

graft harvest.

METHOD DETAILS

Culture methods: Human pluripotent stem cell differentiation in H1 and HUES8 cells was

performed as per the protocol previously published by Hogrebe, et al. 19-9-11 T.H (feeder

independent) induced pluripotent cells were differentiated through stage 4 using the STEMdiff™

Pancreatic Progenitor Kit (#05120) and subsequently as above.

Flow cytometry: Single cell dissociations were fixed in 4% paraformaldehyde and blocked with

5% normal donkey serum in 0.1% Triton-X PBS. Cells were incubated overnight with primary

antibodies, washed and subsequently incubated with appropriate secondary antibodies. Stained,

fixed cells were analyzed by ZE5 cell analyzer (Bio-Rad). Primary and secondary antibodies used:

anti-OCT-3/4 (Santa Cruz Biotechnology, catalog # sc-5279), anti-chromogranin A (Invitrogen,

catalog # PIMA533052), anti-C-peptide (Developmental Studies Hybridoma Bank, University of

Iowa, catalog # GN-ID4-S), anti-NKX6–1 (Developmental Studies Hybridoma Bank, University of

Iowa, catalog # F55A12-S), anti-PDX1 (R&D Systems, catalog #AF2419), AlexaFluor 488

(Invitrogen, catalog # A21208), AlexaFluor 488 (Invitrogen, catalog # A21202), AlexaFluor 647

(Invitrogen, catalog # A31571), AlexaFluor 647 (Invitrogen, catalog # A21447), AlexaFluor 647

(Invitrogen, catalog # A31573).

Static *in vitro* insulin secretion measurements: Static insulin secretion assays on human islets

and SC-islets were performed following 1 h incubations with 1.67 mM and 16.7 mM glucose, and

30mM KCl in HEPES-supplemented KRB buffer containing 135 mM NaCl, 4.7 mM KCl, 1.2 mM

KH_2PO_4, 5 mM $NaHCO_3$, 1.2 mM $MgSO_4.7H_2O$, 1 mM $CaCl_2$, 10 mM HEPES and 0.1% BSA (pH

7.4). Samples were analyzed for insulin release using Human Insulin Chemiluminescence ELISA

(ALPCO; Salem, NH, USA) as per the manufacturer's protocol.

Insulin content of explanted scaffolds: Explanted scaffolds were placed in 9.1 mM HCl 95%

EtOH solution. For extraction, samples were sonicated 5x for 15 sec, resting on ice in between

sonication periods. Samples were rotated at 4°C overnight. Samples were then spun at 2500

RPM for 30 min at 4°C. Supernatants were stored at -80°C until analysis. Samples were analyzed

for insulin release using Human Insulin Chemiluminescence ELISA (ALPCO; Salem, NH, USA)

as per the manufacturer's protocol.

Mitochondrial respirometry: The oxygen consumption rate was measured in stage 6 SC-islets

and primary human islets using the BaroFuse (EnTox Sciences, INC; Mercer Island, WA, USA)

as previously described[20]. 200 islets were placed in Krebs-Ringer Bicarbonate HEPES (KRBH)

buffer with a 2 mM glucose per condition. Islets were subject to the following conditions: 20 mM

glucose, 10 µM oligomycin (Sigma), 1 µM FCCP (Sigma-Aldrich), and 3 mM KCN (Thermo

Scientific) for 60 min each.

Metabolomics: Cells were cultured according to Hogrebe et al[4]. On the day of sample collection,

samples were incubated in MCDB 131 with 10.5 g BSA, 5.2 mL GlutaMAX, 5.2 mL P/S, 5 mg

heparin, 5.2 mL MEM nonessential amino acids (Corning, 20–025-CI), 84 µg ZnSO4

(MilliporeSigma, 10883), 523 µL Trace Elements A (Corning, 25–021-CI), and 523 µL Trace

Elements B (Corning, 25–022-CI) and either 5.5mM or 20mM glucose for 3 hr. Media was

removed, and the cells were washed 1X with ice-cold 150 mM ammonium acetate in LC-MS grade

water. Cells were harvested with 200 µL of ice-cold methanol and frozen at -80°C until sample

preparation. For extraction, 200 µL of cold water was added to the cells and the cells were

scraped. The resulting homogenate was sonicated on ice for 10 sec. 400 µL of chloroform was

added. Samples were centrifuged at 17,000 x g for 10 min and the resulting top layer was

collected and taken to dryness under nitrogen. Samples were reconstituted in 30 µL of 2:1 ACN:

water, filtered and 5 µL were injected for analysis. The samples were separated as previously

described[98]. LC-MS analysis was performed on the Agilent 6546 quadrupole TOF mass

spectrometer, which was coupled to Agilent 1290 LC. The gas temperature was set at 225°C, the

drying gas flow at 10 L/min, the nebulizer at 40 psi, the sheath gas temperature at 300°C, and the

sheath gas flow at 12 L/min. The Fragmentor (Agilent) was set at 125 V, the skimmer at 65 V, and

VCap at 3,000 V. For verification of metabolite identity and retention time, authentic standards of

all measured metabolites were run separately and spiked into pooled samples. Metabolites were

separated on an InfinityLab Poroshell120 HILIC-Z, 2.7 µm, 2.1 × 150 mm column (Agilent

Technologies, CA, USA). Mobile phase A was composed of 90% 10 mM ammonium acetate pH

9.0 with ammonia, 10% acetonitrile (ACN) with 2.5 uM InfinityLab Deactivator Additive (Agilent

Technologies, CA, USA) and mobile phase B was composed of 15% 10 mM ammonium acetate

pH 9.0 with ammonia, 85% acetonitrile with 2.5 uM InfinityLab Deactivator Additive. The flow rate

was 0.25 mL/min, and the gradient was as follows: 0-2 min at 95 % B, 2-5 min at 95% B, 5-5.5

min at 86%, 5.5-8.5 min at 86% B, 8.5-9 min at 84% B, 9-14 min at 84% B, 14-17 min at 80% B,

17-23 min at 60% B, 23-26 min at 60% B, 26-27 min at 95% B, and 27-35 min at 95% B. Column

compartment temperature was kept at 25 °C. Data were acquired in negative mode.

Lipidomics: Media were removed, and metabolites were extracted with solvent (8:1:1 ratio of

methanol/Chloroform/water) containing stable isotope-labeled internal standards. As previously

described[99], an Agilent 6410 triple quadrupole MS system equipped with an Agilent 1200 LC

System and an ESI source was utilized. Metabolite separation was achieved using gradient

elution on a reverse phase XBridge C18 Column (50 mm × 2.1 mm, 2.5 µm, Waters, Milford, MA,

USA) with the corresponding guard column (5 mm x 2.1 mm, 1.7 µm) maintained at 40 °C. LC

vials were maintained at 4 °C in a thermostatic autosampler, and the injection volume was set at

4 µL. The mobile phase consisted of solvent A, 5 mM Ammonium Acetate, and solvent B,

Acetonitrile. The flow rate was 0.25 mL/min. The gradient elution program was as follows: 0–8.50

min, 2% B; 1.5–9 min, linear gradient from 2% to 50% B; 9–14 min, linear gradient from 50% to

95% B; hold at 95% B for 3 min. The flow rate was 300 μL/min. Acylcarnitine species were each

detected by their characteristic LC retention time in the MRM mode following ESI and comparing

relative areas with those of corresponding standards. Concentrations of carnitine, acetylcarnitine

(C2), propionylcarnitine (C3), butyrylcarnitine (C4), isovalerylcarnitine (C5), hexanoylcarnitine

(C6), octanoylcarnitine (C8), myristoylcarnitine (C14), palmitoylcarnitine (C16), and

oleoylcarnitine (C18) were calculated by ratios of peak areas of known concentrations of stable

isotopically-labeled analogs.

mtDNA content and replication assessments: 25 human islets per donor were handpicked and

washed twice with 1X PBS. 100,000 cells from differentiation stage 0 through 6 were washed

twice with 1X PBS. Human islets and stem cells were pelleted and stored at -80°C until DNA

extraction using the Blood/Tissue DNeasy kit (Qiagen; Germantown, MD, USA) as per

manufacturer's instructions. Relative mtDNA content was quantified by qPCR using mtDNA

specific primers ND1 (5'-ACCCCCTTCGACCTTGCCGA-3' and 5'-

GGGCCTGCGGCGTATTCGAT-3'), and nuclear DNA specific primers for GAPDH (5'-

AATCCCATCACCATCTTCCA-3' and 5'-TGGACTCCACGACGTACTCA-3') as previously

described[100]. First-strand mtDNA replication levels were measured by qPCR following digestion

by Mnl I, which site specifically digest double-stranded DNA[101,102]. Briefly, primers specific to 7 S

mtDNA (5'-ATCAACTGCAACTCCAAAGCCACC-3' and 5'-TGATTTCACGGAGGATGGTGGTCA-

3'), whose amplicon contains a Mnl I site, were used to quantify relative single-stranded mtDNA

forming the D-loop. Single-stranded DNA generated from the extension of 7S region was

measured using primers specific to CYTB (5'-AGTCCCACCCTCACACGATTCTTT-3' and 5'-

AGTAAGCCGAGGGCCTCTTTGATT-3'), which also contains an Mnl I site. Primers specific to

COXI (5'-ACCCTAGACCAAACCTACGCCAAA-3' and 5'-TAGGCCGAGAAAGTGTTGTGGGAA-

3'), whose amplicon does not contain an Mnl I site, were used to quantify non-replicating double-

stranded mtDNA. The ratio of 7S or CYTB amplicons to the COXI amplicon were then calculated

after qPCR of MnI I-digested DNA to determine the relative amount of single-stranded mtDNA at

the D-loop strand or initiation of first strand replication, respectively.

Citrate synthase activity assay: 1,000 human islets per donor were handpicked and washed

twice with 1X PBS. One million cells from differentiation stage 0 through 6 were washed twice

with 1X PBS. Human islets and stem cells were lysed with radioimmunoprecipitation assay buffer

containing protease and phosphatase inhibitors (Calbiochem), and insoluble material was

removed by centrifugation. Lysates were used to determine citrate synthase activity with the

MitoCheck Citrate Synthase Activity Assay Kit (Cayman Chemical, Catalog# 701040) per

manufacturer's instructions.

Western blotting: 1,000 human islets per donor were handpicked and washed twice with 1X

PBS. One million cells from differentiation stage 0 through 6 were washed twice with 1X PBS.

Human islets and stem cells were lysed with radioimmunoprecipitation assay buffer containing

protease and phosphatase inhibitors (Calbiochem), and insoluble material was removed by

centrifugation. Equal amounts of proteins were resolved on 4%–20% gradient Tris-glycine gels

(Bio-Rad) and transferred to nitrocellulose membranes (Bio-Rad). Membranes were then blocked

in 5% milk for 1 h and immunoblotting was performed using antibodies to Vinculin (1:1000;

Millipore, Catalog# CP74), TOM20 (1:1000; Cell Signaling Technology, Catalog# 42406), Total

OXPHOS Human Antibody Cocktail (1:1000; Abcam, ab110411), NDUFA9 (1:1000; Abcam;

ab14713), UQCRC2 (1:1000; Abcam; ab14745), PPARα (1:3000; ProteinTech; 66826-1),

phospho-CAMKII (1:1000; Cell Signaling; 12716), CAMKII (1:50; Santa Cruz; sc-5306), phospho-

AMPK (1:1000; Cell Signaling; 2535), AMPK (1:1000; Cell Signaling 2793), phospho-ACC

(1:1000; Cell Signaling; 3661), ACC (1:10,000; ProteinTech; 67373-1) and species-specific HRP-

conjugated secondary antibodies (Vector Laboratories).

RNA-seq: Sequencing was performed by the Advanced Genomics Core at University of Michigan

Medical School. 250 human islets per donor were handpicked and washed twice with 1X PBS.

250,000 cells from differentiation stage 0 through 6 were washed twice with 1X PBS. Total RNA

was isolated from human islet and stem cell samples and DNase treated using commercially

available kits (Omega Biotek and Ambion, respectively). Libraries were constructed and

subsequently subjected to 151 bp paired-end cycles on the NovaSeq-6000 platform (Illumina).

FastQC (v0.11.8) was used to ensure the quality of data. Reads were mapped to the reference

genome GRCm38 (ENSEMBL), using STAR (v2.6.1b) and assigned count estimates to genes

with RSEM (v1.3.1). Alignment options followed ENCODE standards for RNA-seq. FastQC was

used in an additional post-alignment step to ensure that only high-quality data were used for

expression quantitation and differential expression. Data were pre-filtered to remove genes with

0 counts in all samples. Differential gene expression analysis was performed using DESeq2,

using a negative binomial generalized linear model (thresholds: log2(fold change) >0.5 or <-0.5,

FDR (P_{adj}) <0.05). Plots were generated using variations of DESeq2 plotting functions and other

packages with Genialis (Boston, MA, USA).

Pathway analysis of differentially expressed genes: Differentially expressed genes for each

cell-type were identified using DESeq2 packages with Genialis (Boston, MA, USA). The Kyoto

Encyclopedia of Genes and Genomes (KEGG), gene ontology (GO), and WikiPathways

databases via both Enrichr and STRING databases were used to determine association with

particular biological processes, diseases, and molecular functions. The top pathways with an

FDR-adjusted *p*-value < 5% were summarized in the results.

Transcription factor analysis of differentially expressed genes: Differentially expressed

genes for each cell-type were identified using DESeq2 packages with Genialis (Boston, MA,

USA). ChIP Enrichment Analysis (ChEA) and TF Perturbations Followed by Expression

databases via Enrichr were used to determine association with particular biological processes,

diseases, and molecular functions. The top pathways with an FDR-adjusted *p*-value < 5% were

summarized in the results.

Removal of batch effects between differentiations: Bias in gene expression values was

removed using a tool called ComBat-seq, using protocol selection as batch effect and treatment

selection as group effect. ComBat-seq adjusted batch effects in RNA-seq count data. Normalized

RNA-seq counts were uploaded into new collections on the Genialis Expressions. The counts

were adjusted by mapping the quantiles of the empirical distributions of data to the batch-free

distributions[103]. Batch effect adjustment was done using the sva (v3.48.0) R package[104]. The

ComBat-Seq method adjusted raw gene counts for differentiation protocol while retaining

biological signal from treatment. Batch effect corrected raw counts were TPM normalized using

rnanorm (v1.5.1) Python package (https://github.com/genialis/RNAnorm).

Analysis of publicly-available scRNA-seq and scATAC-seq[28]: Pseudo-bulk analyses were

conducted by comparison of log2 (fold change) between SC-β cells compared to primary β cells.

For mapping the SC-β cell differentiation trajectory, we downloaded preprocessed multi-omics

(paired scRNA-seq and scATAC-seq) data from SC islets and performed quality control on each

modality and sample individually and retained only barcodes that passed quality filtering in both

modalities. We embedded the cells using 100 highly variable genes based on the transcriptomic

modality only with phate[104] and clustered the cells based on the first 2 phate components. Based

on marker genes, we identified trajectories from immature to mature endocrine cell types and

extracted the β cell lineage and binned cells along the first phate component. For identification of

primary human β cells, we downloaded preprocessed multi-omics (paired scRNA-seq and

scATAC-seq) data from primary human islets and performed quality control on each modality and

sample individually and retained only barcodes that passed quality filtering in both modalities. We

embedded the cells using UMAP based on the first 20 Harmony[105] corrected principal components

calculated from 2000 highly variable genes based on the transcriptomic modality. We clustered

the cells and based on marker genes, retained only primary human β cells, which we

concatenated with the cells extracted from the SC-β cell differentiation trajectory. For estimating

motif activity, we called peaks based on the ATAC-seq modality with MACS2[106] using the cells

extracted from the SC-β cell differentiation trajectory and counted fragments in peaks for both

primary and SC-β cells using Signac[107]. We identified motifs using the JASPAR2024 database[108],

which we used as input for chromVAR[51].

Transmission electron microscopy: Human islets and SC-islets were handpicked, fixed in 3%

glutaraldehyde and 3% paraformaldehyde in 0.1 M Cacodylate buffer (CB; pH 7.2) at room

temperature for 5 min and then stored at 4 °C. Islets were then washed with PBS and centrifuged

at 2000 rpm for 2 min. Pre-warmed 2% agarose solution was then carefully added to islet pellets,

centrifuged and allowed to cool at 4 °C for 30 min. Samples were then subjected to osmification

in 1.5% $K_4Fe(CN)_6$ + 2% OsO_4 in 0.1 CB for 1 h, dehydrated by serial washes in EtOH (30%,

50%, 70%, 80%, 90%, 95% and 100%) and embedded in Spurr's resin by polymerization at 60 °C

for 24 h. Polymerized resins were then sectioned at 90 nm thickness using a Leica EM UC7

ultramicrotome and imaged at 70 kV using a JEOL 1400 TEM equipped with an AMT CMOS

imaging system. Mitochondrial structures (Aspect Ratio, Perimeter, Circularity) were analyzed and

quantified by ImageJ.

Immunofluorescence of cryosections: SC-islets and scaffolds containing SC-islets were

picked/excised and fixed in 4% paraformaldehyde for 4 h at 4 °C. SC-islets were washed with

PBS and centrifuged at 2000 rpm for 2 min. SC-islets were incubated in methylene blue at room

temperature for 20 min and then washed 2x with PBS. Excised scaffolds were washed 3x with

PBS for 10 min. SC-islets and scaffolds were stored in 30% sucrose in PBS for 16 h at 4 °C . SC-

islets and scaffolds were embedded in OCT and frozen at -80 °C and cryosectioned. SC-islet

sections were rinsed with 75% EtOH for 5 min at RT to remove methylene blue. SC-islet and

scaffold sections were washed 3x with PBS, permeabilized with 0.2% Triton X-100 in PBS for 15

min, washed 3x with PBS and blocked at RT for 1 h with 5% donkey serum in PBT. Primary

antibodies were added for 16 h at 4 °C. Primary antibodies raised against C-peptide (1:50; DSHB,

Catalog#GN-ID4), Glucagon (1:2000; SantaCruz, Catalog# sc-13091), Glucagon (1:500; Cell

Signaling, Catalog#2760), Somatostatin 28 (1:500; Abcam, Catalog#ab111912), Human antigen

clone 235-1 (Millipore Sigma; Catalog#MAB1281), OXPHOS (1:100; Abcam, Catalog#ab110411),

PDX1 (1:1000; Abcam, Catalog#ab47308), NKX6.1 (1:500; Abcam, Catalog#ab221549), and Ki-

67 (1:250, Thermo, Catalog#PA5-19462) were used. Sections were then washed 3x with PBS

and incubated for 2 h at room temperature with species-specific Dylight405, Cy2, Cy3, and Cy5-

conjugated secondary antibodies (Jackson Immunoresearch). Nuclear labelling was performed

using DAPI (Invitrogen). Images were captured under 20X and 60X oil immersion magnification

on an Olympus IX81 microscope. Immunostained islets were captured with Z-stack images and

subjected to deconvolution (CellSens; Olympus). Hormone-positive cells were quantified as

hormone-positive cell/total DAPI-positive (*in vitro*) or Human Ag-positive (post-transplantation)

cells. Ki-67-positive cells were quantified as Ki-67-positive/total C-peptide-positive cells.

Quantification of hormone-positive and Ki-67-positive cells was carried out with blinded samples,

and samples were reidentified after quantification to reduce bias. Quantitative 3-D assessments

of mitochondrial morphology and network were performed on ImageJ using Mitochondria

Analyzer plugin[102,109].

Statistics: In all figures, data are presented as mean ± SEM, and error bars denote SEM, unless

otherwise noted in the legends. Statistical comparisons were performed using unpaired two-tailed

Student's *t*-tests, one-way or two-way ANOVA, followed by Tukey's or Sidak's post-hoc test for

multiple comparisons, as appropriate (GraphPad Prism and R). A P value < 0.05 was considered

significant. For transcriptomic profiling, a FDR-corrected P value <0.05 were considered

significant (see RNA-seq of SC-islets and human islets above). Outlier tests (ROUT method[110])

were routinely performed in GraphPad Prism, and all results were included in analysis unless

designated by the ROUT method.

Data Availability

RNA-seq data have been deposited at GEO (GSE270220). Any additional information required to

reanalyze the data reported in this paper is available from the lead author on request.

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Author Contributions

A.C.L. conceived, designed, and performed experiments, interpreted results, drafted, and

reviewed the manuscript. E.M.W., J.K., K.C., E.B., A.M.S., J.Z., G.L.P, E.L-D., B.H-K., R.K.D.,

J.L., Y.W., E.C.R., M.A., V.S., J.P.P., and L.M. designed and performed experiments and

interpreted results. T.J.H., M.M.C., J.G.S.M., S.P., and L.D.S. designed studies, interpreted

results, and reviewed the manuscript. S.A.S. conceived and designed the studies, interpreted

results, drafted, edited, and reviewed the manuscript.

Conflict of interest statement

S.A.S has received grant funding from Ono Pharmaceutical Co., Ltd. and is a consultant for Novo

Nordisk.

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 FIGURE LEGENDS

Figure 1. SC-islets are transcriptionally and metabolically immature. (A) A schematic

diagram illustrating the cell stages during differentiation of human pluripotent stem cells (hPSCs)

into SC-islets. (B) Differential RNA expression heatmap, displayed as integrated mean $\log_2$ fold

change (FC), from RNA-seq data of representative stage marker expression at each

differentiation stage (n = 4-5 differentiations or independent human islet donors/group). (C)

Immunofluorescence images at 60X magnification of stage 6 SC-islets for C-peptide (green),

Glucagon (red), Somatostatin (red), and DAPI (blue) (n = 3/group). Scale bars, 20 μm. (D)

Differential RNA expression heatmap of genes encoding islet cell maturation markers and

hormones in stage 6 SC-islets (n = 5) and non-diabetic human islet donors (n = 6). FC is relative

to the average across all individual samples. (E) Insulin secretion following static incubations in

1.67 mM glucose (1.67G), 16.7 mM glucose (16.7G), and 30 mM KCl (30 KCl) normalized to

insulin content in stage 6 SC-islets and human islets (n = 3 differentiations or independent human

islet donors/group); *P < 0.05, **P < 0.01 by two-way ANOVA followed by Tukey's multiple

comparisons test. (F) Islet total insulin content normalized to protein content in stage 6 SC-islets

and human islets (n = 5 differentiations and 4 independent human islet donors/group); *P < 0.05

unpaired t-test. Differential metabolomics heatmap demonstrating mean content of pyruvate (G)

or lactate (H) at 5 mM glucose (5G) and 20 mM glucose (20G) for differentiation stages 1 through

6 (n = 3/group) and human islets (n = 5 independent islet donors). (I) Oxygen consumption rate

(OCR) fractional change in stage 6 SC-islets (n = 4 differentiations) and human islets (n = 5

independent islet donors) following exposure to 2 mM glucose (2G), 20 mM glucose (20G), 10

μM oligomycin, 1 μM FCCP, and 3 mM KCN. Data are presented as mean from independent

differentiations or donors $\pm$ SEM; *P < 0.05, **P < 0.01 by unpaired t-test of fractional OCR value

at designated time point for minimal or maximal respiration by oligomycin or FCCP, respectively.

1104

Figure 2. SC-islets display comparable mitochondrial DNA content, mass, and architecture to human islets. (A) Quantification of ratio of mitochondrial DNA to nuclear DNA by qPCR for stage 0-6 cells and human islets (n = 4/group). (B) Quantification of single strand mitochondrial DNA products measured by qPCR for stage 0-6 cells and human islets (n = 3-4/group). (C) Quantification of citrate synthase activity normalized to total protein content for stage 0-6 cells and human islets (n = 3-4/group). (D) Representative Western blot (WB) of TOM20 protein expression for stage 0-6 cells and two different human islet donors (left). Vinculin serves as a loading control. Quantification (right) of TOM20 expression (normalized to Vinculin expression) via densitometry of the Western blots (n = 4-5/group). (E) Representative transmission electron microscopy (TEM) images of stage 6 SC-β cells and primary human β cells (left) and quantification of mitochondria ultrastructure (right) (n = 3/group, ~150 β cell mitochondria scored/sample).

Figure 3. Transcriptomic profiling of SC-islets versus human islets reveals potential limitations in mitochondrial metabolism. (A) Volcano plot depicting differential RNA expression in stage 6 SC-islets compared to human islets (n = 6 differentiations or independent human islet donors/group). Significantly differentially expressed genes demarcated by -FDR > 0.5 and |FC| > 1.4. MitoCarta 3.0 genes are highlighted in purple. (B) Differential RNA expression heatmaps from RNA-seq data demonstrating expression of significantly differentially expressed MitoCarta 3.0 genes in stage 0-6 cells compared to human islets (n = 4-6 differentiations or independent human islet donors/group). FC is relative to the average across all individual samples. Selected terms from Enrichr analysis of differentially expressed MitoCarta genes from stage 6 SC-islets vs. human islets analyzed with the following libraries: (C) WikiPathway 2023 Human and (D) KEGG 2021 Human. (E) STRING analysis of MitoCarta genes with reduced expression in stage 6 SC-islets compared to human islets.

Figure 4. Reduced expression of the OXPHOS machinery in SC-islets. (A) Differential RNA

expression heatmap, displayed as integrated mean $\log_2$ fold change (FC), from RNA-seq data of

genes related to glucose transport and glycolysis at each differentiation stage and human islets

(*n* = 4-5 differentiations or independent human islet donors/group). (B) Differential RNA

expression heatmap (presented as integrated mean/group) of genes encoding OXPHOS

components at each differentiation stage as well as human islet donors (*n* = 4-6/group). (C)

Differential RNA expression heatmaps from RNAseq demonstrating expression of OXPHOS

machinery genes in stage 0-6 cells and human islets (*n* = 4-6 differentiations or independent

human islet donors/group). FC is relative to the average across all individual samples. (D)

Differential RNA expression heatmap, displayed as integrated mean $\log_2$FC, from scRNA-seq of

genes related to glycolysis and glucose transport and OXPHOS between SC-β cells to primary β

cells as well as OXPHOS genes between SC-α cells to primary α cells. Data curated from

Augsornworawat *et al.*[28]. (E) Representative Western blot analysis of expression of OXPHOS

subunits, including ATP5A (Complex V), UQCRC2 (Complex III), NDUFA9 (Complex I), SDHB

(Complex II), and COXII (Complex IV) in stage 0-6 cells and human islets (left). Vinculin serves

as a loading control. Quantification (right) of ATP5A, UQCRC2, NDUFA9, SDHB, and COXII

protein expression by densitometry (normalized for Vinculin expression) of the Western blots of

stage 0-6 cells and human islets (*n* = 4 differentiations or independent human islet donors/group);

*$P < 0.05$, **$P < 0.01$, ***$P < 0.001$, by one-way ANOVA followed by Dunnett's multiple

comparisons test.

1150

Figure 5. Limitations in mitochondrial lipid metabolism in SC-islets. Differential RNA

expression heatmaps from RNA-seq data demonstrating expression of fatty acid biosynthesis (A)

or fatty acid β-oxidation (B) genes in stage 0-6 cells and human islets (*n* = 4-6 differentiations or

independent human islet donors/group). FC is relative to the average across all individual

samples. (C) Differential RNA expression heatmaps from RNA-seq data demonstrating

expression of lipid metabolism genes in stage 0-6 cells and human islets (n = 3-5 differentiations

or independent human islet donors/group). FC is relative to the average across all individual

samples. Lipidomics profiling of (D) L-carnitine (E) and short- or long-chain acylcarnitines

(normalized to protein content) for stage 1-6 cells (n = 3/group) and human islets (n = 4-5

independent islet donors); *P < 0.05, **P < 0.01, ****P < 0.0001 by one-way ANOVA followed by

Dunnett's multiple comparisons test. (F) Differential RNA expression heatmaps from RNA-seq

data demonstrating expression of expression of carnitine biosynthesis and short- and medium-

chain acylcarnitine synthesis genes in stage 0-6 cells and human islets (n = 3-5 differentiations

or independent human islet donors/group). FC is relative to the average across all individual

samples.

1166

Figure 6. SC-islets display reduced expression of PPARα and PPARγ transcriptional

targets. Enrichr analysis of significantly differentially expressed genes from stage 6 SC-islets

compared to human islets analyzed with the following libraries: (A) ChEA 2022 and (B)

Transcription factor (TF) perturbations followed by expression. (C) Differential RNA expression

heatmap from RNA-seq data demonstrating integrated mean expression of PPARα targets for

stage 6 SC-islets and human islets (n = 6/group). (D) Differential RNA expression heatmap from

RNA-seq data demonstrating integrated mean expression of genes encoding components of the

PPAR-RXR transcriptional binding complex, including transcriptional co-activators, in stage 6 SC-

islets and human islets (n = 6/group). (E) Representative Western blot analysis of expression of

PPARα in stage 0-6 cells and human islets (left). Vinculin serves as a loading control.

Quantification (right) of PPARα protein expression by densitometry (normalized to Vinculin

expression) of the Western blots of stage 0-6 cells and human islets (n = 3 differentiations or 4

independent human islet donors/group); *P < 0.05 by one-way ANOVA followed by Dunnett's

multiple comparisons test. (F) Differential RNA expression heatmap generated from publicly-

available scRNA-seq data[28] demonstrating integrated mean expression of genes encoding

components of the PPAR-RXR transcriptional binding complex, including transcriptional co-

activators, in SC-β cells and primary β cells. (G) Differential chromatin accessibility heatmap

generated from publicly-available scATACseq data[28] demonstrating integrated mean motif

accessibility of PPAR and RXR binding sites in SC-β cells and primary β cells. (H) Minimum to

maximum box and whisker plots of normalized expression of *PPARA* across SC-β cells

differentiation trajectory and primary β cells generated from publicly-available scRNA-seq data[28]

(*n* = 3-4 differentiations or independent human islet donors/group). (I) Minimum to maximum box

and whisker plots of PPARA module scores across SC-β cells differentiation trajectory and primary

β cells generated from publicly-available scRNA-seq data[28] representing PPARA target gene

expression (*n* = 3-4 differentiations or independent human islet donors/group). (J) Minimum to

maximum box and whisker plots of motif deviation z-scored across SC-β cells differentiation

trajectory and primary β cells generated from publicly-available scATAC-seq data[28] representing

PPARα motif activity (*n* = 3-4 differentiations or independent human islet donors/group).

1195

Figure 7. WY14643 promotes the formation of SC-β cells. (A) A schematic diagram illustrating

use of the PPARα agonist WY14643 within the differentiation protocol. (B) Immunofluorescence

images at 60X magnification (left) of vehicle control and WY14643-treated stage 6 SC-islets for

C-peptide (green), glucagon (red), and DAPI (DNA - blue) or C-peptide (green), somatostatin

(red), and DAPI (DNA - blue) (*n* = 3/group). Scale bars, 20 µm. Quantification (right) of C-peptide

positive cells, dual hormone-positive cells, either C-peptide and glucagon (GCG) dual-positive or

C-peptide and somatostatin (SST) dual-positive, as well as GCG-positive cells, and SST-positive

cells (*n* = 3 differentiations/group, ~1000 cells quantified/differentiation and respective condition);

***P* < 0.001 by unpaired Student's *t*-test. (C) Immunofluorescence images at 20X magnification

(left) of vehicle control and WY14643-treated stage 6 SC-islets for C-peptide (green) and NKX6.1

(red) (*n* = 3/group). Scale bars, 50 µm. Quantification (right) of the fractional change of NKX6.1/C-

peptide co-positive cells per high powered field (*n* = 12-15 fields quantified/group, ~1000 cells

quantified/differentiation and respective condition from 3 differentiations/group); **P < 0.01 by

unpaired Student's *t*-test. (D) Immunofluorescence images at 20X magnification (left) of vehicle

control and WY14643-treated stage 6 SC-islets for C-peptide (green) and PDX1 (red) (n =

3/group). Scale bars, 50 µm. Quantification (right) of the fractional change of PDX1/C-peptide co-

positive cells per high powered field (n = 12-15 fields quantified/group, ~1000 cells

quantified/differentiation and respective condition from 3 differentiations/group). (E) Differential

RNA expression heatmap (presented as integrated mean/group) of PPARα targets for vehicle

control and WY14643-treated stage 6 SC-islets (n = 4/group). (F) Differential RNA expression

heatmap (presented as integrated mean/group) of mature β cell markers for vehicle control and

WY14643-treated stage 6 SC-islets (n = 4/group). (G) Differential RNA expression heatmap

(presented as integrated mean/group) of immature β cell markers for vehicle control and

WY14643-treated stage 6 SC-islets (n = 4/group). (H) Differential RNA expression heatmap

(presented as integrated mean/group) of genes encoding OXPHOS components for vehicle

control and WY14643-treated stage 6 SC-islets (n = 4/group). (I) Differential RNA expression

heatmap (presented as integrated mean/group) of genes encoding glucose transporters and

glycolytic enzymes for vehicle control and WY14643-treated stage 6 SC-islets (n = 4/group). (J)

Insulin secretion (left) with insulin stimulation index (right) following static incubations in 1.67 mM

glucose (1.67G), 16.7 mM glucose (16.7G) and 30 mM KCl (30KCl) normalized to insulin content

in vehicle control and WY14643-treated stage 6 SC-islets (n = 4 differentiations/group); **P < 0.01

by two-way ANOVA. (K) Oxygen consumption rate (OCR) fractional change in vehicle control and

WY14643-treated stage 6 SC-islets (n = 3 differentiations) following exposure to 2 mM glucose

(2G), 20 mM glucose (20G), 10 µM oligomycin, 1 µM FCCP, and 3 mM KCN. Data are presented

as mean from independent differentiations $\pm$ SEM; P = 0.057 by ANOVA of fractional OCR value

at designated time points following 20 mM glucose exposure. (L) Representative Western blot

analysis of NDUFA9 protein expression for vehicle control and WY14643-treated stage 6 SC-

islets (left). Vinculin serves as a loading control. Quantification (right) of NDUFA9 expression

(normalized to Vinculin expression) via densitometry of the Western blots (n = 4/group); *P < 0.05

by unpaired Student's t-test.

1236

Figure 8. Formation of SC-β cells by WY14643 is durable *in vivo* following transplantation.

(A) A schematic diagram illustrating the addition of WY14643 to the differentiation protocol,

seeding of SC-islets onto poly(lactide-*co*-glycolide) (PLG) microporous scaffolds, and scaffold

transplantation into NSG mice. (B) Immunofluorescence images at 20X and 60X magnification of

stage 6 SC-islets for C-peptide (green), human nuclear antigen (red), and DAPI (blue) (n =

3/group). Human nuclear antigen was used to distinguish human transplanted cells from murine

recipient cells in the inguinal fat pad. Scale bars, 20 μm. (C) Immunofluorescence images at 60X

magnification of vehicle control and WY14643-treated stage 6 SC-islets in explanted scaffold from

NSG mice 10 days post-transplant for C-peptide (green), glucagon (red), and DAPI (blue) or C-

peptide (green), somatostatin (red), and DAPI (blue) (n = 3-4/group). Scale bars = 20 μm. (D)

Quantification of C-peptide-positive cells only, dual hormone-positive cells, both C-peptide- and

GCG- or SST-positive cells, GCG-positive cells, and SST-positive cells by immunostaining in

vehicle control and WY14643-treated stage 6 SC-islets in explanted scaffold from NSG mice 10

days post-transplant (n = 3-4/group); *P < 0.05 by unpaired Student's t-test. (E) Insulin content

normalized to total protein content for vehicle control and WY14643-treated explanted scaffolds

from NSG mice 10 days post-transplant (n = 3-4/group); *P < 0.05 by unpaired Student's t-test.

(F) Quantification of Ki-67 and C-peptide co-positive cells by immunostaining in vehicle control

and WY14643-treated stage 6 SC-islets in explanted scaffold from NSG mice 10 days post-

transplant (n = 4/group). (G) Immunofluorescence images (right) at 60X magnification of vehicle

control and WY14643-treated stage 6 SC-islets in explanted scaffold from NSG mice 10 days

post-transplant for C-peptide (green), OXPHOS (red), and DAPI (blue) (n = 3-4/group). Scale

bars, 20 μm. Quantification (right) of mitochondrial morphology and networking utilizing

Mitochondria Analyzer (n = 3-4/group, ~150 cells analyzed per differentiation and treatment).

A

-log10(FDR)

log2(FC) SC-islets vs. human islets

B

MitoCarta 3.0 Genes

log2(FC)

Differentiation Stage

0 1 2 3 4 5 6 Human Islets

C

WikiPathway 2023 Human

- **TCA cycle**
- **Mitochondrial complex I assembly**
- **Mitochondrial FA oxidation disorders**
- Urea cycle & related disorders
- **Mitochondrial long chain FA β-oxidation**
- Riboflavin & CoQ disorders
- **Electron transport chain/OXPHOS**
- **FA β-oxidation**
- Amino acid metabolism
- **PPAR signaling pathway**

-log10(FDR)
10
8
6
4

Number of Genes
• 4
• 8
• 20

Combined Score
0   200   400   600

D

KEGG 2021 Human

- Valine, leucine, & isoleucine degradation
- Glycine, serine, & threonine degradation
- **Fatty acid degradation**
- **TCA cycle**
- Arginine & proline metabolism
- **Pyruvate metabolism**
- **Oxidative phosphorylation**
- **PPAR signaling**
- Diabetic cardiomyopathy
- Parkinson's disease

-log10(FDR)
10
8
6

Number of Genes
• 5
• 10
• 20

Combined Score
0   200   400   600   800

E

Biologic Process (Gene Ontology)	FDR
Mitochondrial Translation (GO:0032543)	1.83E-18
Oxidative Phosphorylation (GO:0006119)	9.11E-18
Cellular Lipid Catabolic Process (GO:0044242)	1.50E-05
Fatty Acid Biosynthetic Process (GO:0006633)	5.40E-03
Carnitine Metabolic Process (GO:0009437)	6.10E-03