CRISPR Screening Uncovers a Long-Range Enhancer for *ONECUT1* in Pancreatic Differentiation and Links a Diabetes Risk Variant

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Abstract

Functional enhancer annotation is a valuable first step for understanding tissue-specific

transcriptional regulation and prioritizing disease-associated non-coding variants for

investigation. However, unbiased enhancer discovery in physiologically relevant contexts

remains a major challenge. To discover regulatory elements pertinent to diabetes, we conducted

a CRISPR interference (CRISPRi) screen in the human pluripotent stem cell (hPSC) pancreatic

differentiation system. Among the enhancers uncovered, we focused on a long-range enhancer

~664 kb from the *ONECUT1* promoter, as coding mutations in *ONECUT1* cause pancreatic

hypoplasia and neonatal diabetes. Homozygous enhancer deletion in hPSCs was associated with

a near-complete loss of *ONECUT1* gene expression and compromised pancreatic differentiation.

This enhancer contains a confidently fine-mapped type 2 diabetes (T2D) associated variant

(rs528350911) which disrupts a GATA motif. Introduction of the risk variant into hPSCs

revealed substantially reduced binding of key pancreatic transcription factors (GATA4, GATA6

and FOXA2) on the edited allele, accompanied by a slight reduction of *ONECUT1* transcription,

supporting a causal role for this risk variant in metabolic disease. This work expands our

knowledge about transcriptional regulation in pancreatic development through the

characterization of a long-range enhancer and highlights the utility of enhancer discovery in

disease-relevant settings for understanding monogenic and complex disease.

Main

A spectrum of genetic causality often influences the onset and varying severity of symptoms for a disease[1]. This concept is exemplified in metabolic conditions like diabetes, where a single nucleotide alteration in a protein coding region can cause severe neonatal diabetes mellitus (NDM), while the aggregation of polygenic effects is thought to confer susceptibility to type 2 diabetes (T2D)[2,3]. However, most genetic variants have moderate effects and are in unannotated non-coding genomic regions, with the affected genes typically unknown[3,4]. Therefore, there is a pressing need to understand the functional impact of disease-associated variants uncovered by genome-wide association studies (GWAS) and manage the burgeoning set of variants of unknown significance identified through the widespread adoption of whole genome sequencing.

It is hypothesized that many genetic variants affect enhancers[5], and alterations in their sequences can affect gene expression, contributing to both Mendelian and complex disease traits. A clear illustration of this impact is shown by a mutation in a *SHH* enhancer, ~1 Mb from the gene promoter, causing limb malformation[6]. This represents one of a handful of clinical genetics examples where enhancer mutations have a pronounced pathological impact[7]. While the specific functional consequences of individual sequence alterations in common complex diseases are generally unknown, a significant proportion of disease-associated genomic variations identified through GWAS are located within putative enhancers[5]. Together, these findings have contributed to the inception of the enhanceropathy concept[7]. However, diagnosis of suspected enhanceropathies remains difficult due to the lack of functional annotation of most enhancers. While putative enhancers can be identified based on genomic features like chromatin accessibility, achieving scalable functional characterization in tissue-relevant settings poses a

considerable challenge. One obstacle is the variable distance that enhancers can have from their

target genes. In our recent interrogation of developmental enhancers, we found that 41% of the

enhancer identified were located >100 kb away from the target gene promoter[8], yet practical

considerations often lead many functional enhancer screens to confine their search to regions

within 100 kb of the target gene promoter[9]. Consequently, the overall prevalence of long-

distance enhancer-promoter regulations remains unclear despite a growing set of long-range

enhancers characterized through genetic studies[10].

Human pluripotent stem cell (hPSC) differentiation represents a powerful system for disease

modeling. The integration of genetic perturbation and genomic approaches has proven

instrumental for investigating the cascade of gene and enhancer activation during development,

holding promise for understanding long-term health implications. We and others have shown

neonatal diabetes (NDM) gene knockout in hPSCs recapitulates key aspects of human genetic

conditions[11]. In addition, the hPSC platform has been used to study the impact of clinically

observed recessive mutations in a *PTF1A* enhancer, shedding light on its role in pancreatic

agenesis and NDM[12,13]. However, many cases of NDM lack a known genetic cause even with

whole exome sequencing, underscoring the importance of identifying developmental

enhancers[14]. Significant progress in biochemical characterization has narrowed the search space

for enhancers, but many putative enhancers have not been functionally interrogated.

Furthermore, the extensive epigenetic rewiring during development further swells the number of

potential enhancers, adding complexity to the characterization task[15,16]. This challenge is

becoming more tractable with the use of CRISPR-Cas9 and dCas9-based CRISPR interference

(CRISPRi) tools, enabling the simultaneous perturbation of genes or enhancers en masse[9]. We

and others have applied CRISPR-Cas9 in hPSCs to uncover genes important for gastrulation and

pancreatic development[17-22]. Recent CRISPRi screens have further identified enhancers involved

in hPSC endoderm and cardiac differentiation[8,23,24]. These advances motivated us to leverage our

epigenetic characterization of the hPSC pancreatic differentiation system[25] to uncover enhancers

with potential roles in diabetes.

The hPSC pancreatic differentiation system directs cells progressively through the definitive

endoderm (DE), gut tube (GT), and posterior foregut (PFG, also known as PP1) stages that

model embryonic development in humans[26]. Building on our recent success in using a cell

identity gene as a readout for enhancer discovery in cell state transitions[8], we utilized the

developmental initiation of diabetes-associated gene *PDX1* expression at the PFG stage as a

readout for the onset of pancreatic differentiation[11,17]. We then selected candidate transcription

factors (TFs) likely to influence PDX1+ identity commitment and hypothesized that repressing

enhancers of these TFs would affect PDX1 expression. Our subsequent screen indeed uncovered

enhancers of multiple TF genes that affected PDX1 activation during pancreatic specification.

We focused on a distal enhancer of *ONECUT1*, located ~664 kb from the gene promoter. This

choice was motivated by the presence of a high-confidence fine-mapped T2D-associated single

nucleotide polymorphism (SNP) within this enhancer[27] and clinical genetics findings indicating

that coding mutations in *ONECUT1* cause NDM[28,29]. We show that *ONECUT1* enhancer deletion

in hPSCs resulted in an almost complete absence of *ONECUT1* transcript, protein, along with a

reduced proportion of cells expressing PDX1 during pancreatic differentiation. Enhancer

deletion had locus-wide effects within the topologically associating domain (TAD), with reduced

anti-sense transcription and decreased H3K27ac levels. Finally, we generated hPSC lines

harboring the T2D variant (rs528350911) and found the risk variant significantly impaired the

binding of key pancreatic transcription factors GATA4, GATA6, and FOXA2, but had a smaller

effect on *ONECUT1* transcription compared to enhancer knockout. Our study not only identified

an essential enhancer of *ONECUT1* in pancreatic development but also developed and reduced to

practice a framework for prioritizing disease-associated variants for investigation.

Results

CRISPRi enhancer screen design

To discover enhancers involved in pancreatic development and diabetes, we conducted a

CRISPRi screen in hPSC pancreatic differentiation, using PDX1 expression as a readout. PDX1

is a key marker for the initiation of pancreatic differentiation and is indispensable for mammalian

pancreatic development[30,31]. Homozygous and heterozygous mutations in *PDX1* cause NDM and

maturity onset diabetes of the young (MODY), respectively[32,33]. Additionally, genomic variants

in the *PDX1* locus are associated with elevated T2D risk[34]. To identify putative enhancer regions

for study, we first selected 23 candidate TFs (including PDX1) based on their likely impact on

PDX1+ induction and diabetes by integrating findings from murine experiments[30,31,35-53], clinical

genetics, and our previous CRISPR-Cas9 screens in hPSC pancreatic differentiation (Fig. 1B).

11 of these TFs have known roles in human metabolic disorders or NDM[29,32,54-59]. 14 TFs,

including GATA6, RFX6, and ONECUT1, have also been shown to affect pancreatic

differentiation and PDX1 expression in hPSC disease models or CRISPR-Cas9 coding screens

for regulators of DE and PFG specification[17,18,28,60,61].

TADs have been proposed to play a key role in enhancer function[62], so we conducted Hi-C

assays on PFG stage cells and used TAD information to delineate a ~2-6 Mb linear window

around each TF for interrogation (Sup. Table 1). Within each window, we collated all accessible

regions identified based on ATAC-seq peaks across the DE, GT, and PFG stages for

interrogation (Fig. 1C, Sup. Table 1)[25]. These accessible regions also included 758 annotated

promoters, facilitating the benchmarking of CRISPRi screen results against our previous

genome-scale CRISPR-Cas9 coding screen conducted in the same differentiation context[17].

Thus, we designed a library with 37,184 gRNAs targeting 6,443 regions (avg. length of 300 bp),

and 1100 safe-targeting[63] and 463 non-targeting gRNAs[64], and cloned the sequences into a

lentiviral backbone (Sup. Table 1).

CRISPRi screen identifies enhancers and genes necessary for human pancreatic

differentiation

We conducted a pooled CRISPRi screen in our hPSC pancreatic differentiation system[17], with

two independent infections and differentiations referred to as Screen 1 and 2 (Fig. 1A). For each

screen, we infected inducible dCas9-KRAB H1 hPSCs[65] with the gRNA library and induced

dCas9-KRAB expression. Following differentiation, we enriched for PDX1+ and PDX1- cells at

the PFG stage through fluorescence-activated cell sorting (FACS) (Fig. 1D, Sup. Fig. 1A).

Subsequently, next-generation sequencing determined gRNA abundance within the sorted

populations, and MAGeCK analysis[66] led us to prioritize 36 top hits causing a PDX1 decrease

(Fig. 1E, Sup. Fig 1B, Sup. Table 1). Both promoter (19) and non-promoter (17) hits have greater

sequence conservation compared to all regions investigated, as shown by the average

conservation scores calculated from PhyloP100[67] (Sup. Fig. 1C). Furthermore, we examined all

interrogated promoter regions and found the gene hits to be largely concordant with our previous

Cas9 coding screen conducted in a similar pancreatic differentiation context[17] (Fig. 1F). These

findings support the utility of CRISPRi repression for the discovery of both coding and non-

coding regulators of hPSC differentiation.

We next focused on non-promoter hits and examined H3K27ac, a histone mark associated with

active enhancers[68]. Non-promoter hit regions did not have H3K27ac peaks in undifferentiated

hPSCs, but gained H3K27ac at various stages during pancreatic differentiation (Fig. 1G,H),

indicating stage-specific enhancer activity. We further reasoned that each enhancer hit is likely to

correspond to a promoter hit for the regulated gene. Indeed, almost all discovered enhancers

were linked to a promoter hit within the linear neighborhood, resulting in a total of 15 enhancer-

gene pairs (Fig. 1I, Sup. Fig. 1D). This approach supports the specificity of the enhancer-gene

regulations. In addition, we observed that the enhancer-promoter distances ranged from 2 to 664

kb, with three enhancers found beyond 100 kb from the transcription start site (TSS) (Fig. 1J).

Thus, unbiased screening in hPSC pancreatic differentiation has uncovered many previously

undiscovered enhancers, expanding our understanding of gene regulation during human pancreas

development.

***ONECUT1e-664kb* deletion causes allele-specific loss of *ONECUT1* expression**

The CRISPRi screen identified the *ONECUT1* promoter as well as a site ~664 kb from the

ONECUT1 TSS, denoted as *ONECUT1e-664kb* henceforth. We observed that at the PFG stage,

the *ONECUT1* promoter showed significantly increased Hi-C contact with a region

encompassing *ONECUT1e-664kb* compared to the DE stage (Fig. 2A, 2B). This observation

supports 3D enhancer-promoter interaction and suggests substantial enhancer rewiring during

differentiation. We chose *ONECUT1e-664kb* for further investigation because *ONECUT1*

mutations cause pancreatic agenesis in humans[28,29] and *ONECUT1* is necessary for efficient

hPSC differentiation to pancreatic progenitors[28]. Consistent with the clinical findings, *ONECUT1*

null mice had a reduced number of endodermal cells expressing PDX1 at pancreatic budding[53]

accompanied by compromised pancreatic organogenesis[69-71]. To validate the regulatory impact

of *ONECUT1e-664kb* on *ONECUT1*, we used CRISPR-Cas9 deletion as an orthogonal method

to complement our dCas9-KRAB based repression approach. Using paired gRNAs flanking a

~750 bp region, we generated two heterozygous and one homozygous clonal enhancer deletion

lines (Fig. 2C). *ONECUT1* transcription is activated during the GT to PFG transition (Sup. Fig.

2A), so we differentiated the homozygous and heterozygous enhancer knockout (eKO) lines to

the PFG stage and assessed PDX1 and ONECUT1 protein levels via flow cytometry (Fig. 2D).

Homozygous eKO led to a near-complete depletion of ONECUT1+ cells (Fig. 2E). In addition,

the proportion of cells activating PDX1 expression was significantly lower in homozygous eKO

cells, showing that the loss of ONECUT1 impaired pancreatic differentiation (Fig. 2F). These

findings were further confirmed through immunofluorescence staining for PDX1 and ONECUT1

(Sup. Fig. 2B). While heterozygous lines did not exhibit a significantly lower percentage of

PDX1+ or ONECUT1+ cells, there was a significant decrease in the mean fluorescence intensity

signal of ONECUT1 among the PDX1+ cells in heterozygous cells compared to wild-type (WT)

cells (Fig. 2G). This suggested that the heterozygous cells had lower ONECUT1 expression

compared to WT cells.

To directly assess the *cis* effect of heterozygous eKO on *ONECUT1* transcription, we pursued a

more quantitative approach and designed an allele-specific droplet digital PCR (ddPCR) assay

for the *ONECUT1* transcript. Given the substantial linear genomic distance between

ONECUT1e-664kb and the *ONECUT1* gene, we first called and phased all SNPs in the

ONECUT1 locus after conducting targeted long-read Nanopore sequencing with adaptive

sampling (Sup. Fig. 2D). This approach allowed us to identify a SNP in the 3'UTR of *ONECUT1*

for allele-specific ddPCR (Fig. 2H). We further identified a genotyping SNP close to

ONECUT1e-664kb for phasing the deleted enhancer in the heterozygous lines via PCR (Fig. 2I,

Sup. Fig. 2D). Both heterozygous lines had enhancer deletions on the same allele (designated as

"Allele 2") (Fig. 2J). ddPCR assays performed on RNA extracted from cells differentiated to the

PFG stage showed that homozygous eKO lines lost >95% of transcription from both alleles (Fig.

2K). Heterozygous lines exhibited a specific loss of transcription from Allele 2, resulting in a

significantly skewed ratio of transcripts from Allele 2 over Allele 1. This skewed ratio was not

observed in control ddPCR assays conducted on genomic DNA (gDNA) (Sup. Fig. 2C). These

findings demonstrate that the transcriptional impact of *ONECUT1e-664kb* deletion on

ONECUT1 expression is due to direct in cis regulation, rather than indirect effect through the

downregulation of the ONECUT1 TF or other mechanisms. Notably, one of the heterozygous

eKO lines (Het. 1) had the enhancer sequence inverted on Allele 1, and *ONECUT1* transcription

from Allele 1 was not significantly affected by this inversion. Together, these results establish

ONECUT1e-664kb as a critical *cis*-regulatory element for *ONECUT1* activation in pancreatic

development.

***ONECUT1* enhancer deletion causes transcriptional and epigenomic changes in the locus**

The large impact of *ONECUT1e-664kb* deletion on *ONECUT1* transcription led us to investigate

its potential effects on transcription at the broader genomic locus. We performed TruSeq

Stranded Total (Ribodepletion) RNA-seq on WT, heterozygous, and homozygous eKO cells

differentiated to the GT and PFG stages. Consistent with our ddPCR results, homozygous eKO

cells had an approximately 30-fold decrease in *ONECUT1* reads at the PFG stage compared to

reads in WT cells. Some enhancers may serve as "hubs" that simultaneously regulate the

expression of multiple genes in a locus[72]. While we did not detect any significant transcription

changes of genes in the adjacent TADs, we did observe a statistically significant decrease in

WDR72 expression in homozygous *ONECUT1e-664kb* cells (Fig. 3A). Notably, *WDR72* is the

only other protein-coding gene in the same TAD as *ONECUT1*. However, ddPCR assays did not

show a statistically significant difference in *WDR72* transcripts, despite a noticeable trend

towards a decrease (Sup. Fig. 3A), which may be attributed to the low expression of *WDR72* in

the PFG stage. In summary, we conclude that *ONECUT1e-664kb* primarily regulates *ONECUT1*

transcription during pancreatic differentiation.

We further investigated the impact of *ONECUT1e-664kb* deletion on *ONECUT1* promoter

antisense (PAS) RNA transcription. Transcription-associated PAS RNAs have been extensively

documented[73,74], and antisense RNAs can be quantified in strand-specific RNA-seq datasets[75].

Indeed, we observed an adjacent non-coding region antisense to *ONECUT1* that had lower PAS

transcription in homozygous eKO cells compared to WT cells (Fig. 3B). Quantifying antisense

transcripts in this region across three differentiation replicates showed that during the GT-PFG

transition, *ONECUT1* transcriptional activation in WT cells was associated with a significant

increase in *ONECUT1* PAS transcripts. These transcripts were significantly reduced in both

heterozygous and homozygous eKO lines at the PFG stage (Fig. 3C). In contrast, sense

transcripts across the same non-coding region showed no change during the GT-PFG transition

and were unperturbed upon enhancer deletion (Sup. Fig. 3B). Thus, enhancer deletion has

transcriptional effects beyond the decrease of the *ONECUT1* transcript, illustrating an additional

transcriptional phenotype of enhancers loss.

H3K27ac is an epigenetic mark associated with active enhancers and transcription. A previous

study documented that *PTF1A* enhancer deletion caused H3K27ac loss at the *PTF1A* promoter as

well as adjacent regions in the genomic locus[12]. However, given that *PTF1A* encodes a TF, it is

challenging to distinguish *trans* and *cis* effects of enhancer deletion. We reasoned that analyzing

allele-specific H3K27ac ChIP-seq reads could aid in examining *cis*-consequences of enhancer

deletion, as a *cis*-effect would manifest as an imbalance in allele-specific reads within

heterozygous eKO cells. We performed H3K27ac ChIP-seq on PFG cells differentiated from

WT, heterozygous, and homozygous eKO lines. First, we compared WT and homozygous eKO

cells. The decrease in H3K27ac in the *ONECUT1* TAD was the most pronounced of all regions

on chromosome 15 (Sup. Fig. 3C), strongly suggesting cis effects. Among the 16 regions with

the most significant H3K27ac reduction (adjusted p-value < 1E-15) (Fig. 3D), 10 contained

heterozygous SNPs. These SNPs allowed us to determine changes occurring *in cis* with enhancer

deletion by calculating the allele-specific read ratio in both WT and heterozygous eKO cells

(Sup. Fig. 3D). As controls, we identified regions with heterozygous SNPs but did not meet the

significance threshold. For all top differential peaks, H3K27ac was decreased in cis to the allele

with enhancer deletion in heterozygous eKO cells, with some positions losing nearly all

H3K27ac (Fig. 3E,3F), whereas control regions showed a near 1:1 allele ratio in both WT and

heterozygous eKO cells. This demonstrates that *ONECUT1e-664kb* controls the activity of

multiple enhancers within the locus, arguing for direct crosstalk among the cis-regulatory

elements. In addition, the reduction of PAS transcripts and H3K27ac signals in *cis* with enhancer

deletion also illustrate the impact of enhancer dysregulation on genome function beyond gene

transcription.

T2D-associated SNP decreases pancreatic TF recruitment to *ONECUT1e-664kb*

To explore the potential impact of *ONECUT1e-664kb* in clinical genetics, we examined SNP

metabolic disease-association statistics in the *ONECUT1* locus using data from a recent T2D

GWAS meta-analysis[27] and other studies[76,77]. We observed one distinct T2D disease association

signal around the *ONECUT1* gene with a largely unresolved 99% credible set spanning ~100kb

genomic region (hg38, chr15:52777944-52873484). We identified a second independent T2D

disease-causal signal 664 kb away from *ONECUT1* and close to *WDR72*, that is driven by a

single variant, rs528350911 (G/C, reverse strand, minor allele frequency = 0.0068)[27] (Fig 4A).

This variant is directly within *ONECUT1e-664kb* and is highly confidently fine-mapped with the

posterior probability of association (PPA) close to 1 (0.994)[27]. To predict if rs528350911 would

affect enhancer activation in pancreatic development, we trained a ChromBPNet[78,79] model on

PFG-stage ATAC-seq data[25]. Then, we computed per nucleotide DeepLift[80] attribution scores

across the region for both the reference and alternate sequences of rs528350911. The decreased

attribution for an entire GATA motif suggests that the G in the GATA motif is crucial for the

model's genomic accessibility prediction, and the alternate sequence associated with T2D risk

leads to an attribution loss at the site (Fig. 4B).

GATA4 is known for its important role in pancreatic differentiation[60] and is linked to neonatal

diabetes[81]. At the PFG stage, GATA4, along with key pancreatic TFs GATA6 and FOXA2,

binds to the *ONECUT1e-664kb* enhancer region (Sup. Fig. 4A)[25,60]. This motivated us to

investigate the molecular impact of rs528350911. We engineered three heterozygous hPSC lines

with the risk variant: two carried the variant on Allele 2, and one on Allele 1 (Fig. 4C). While all

lines differentiated to the PFG stage had a similar proportion of ONECUT1 and PDX1

expressing cells (Fig. 4D, Sup. Fig. 4B), we observed a significant reduction in ONECUT1 MFI

for two of the three lines (Fig. 4E). Allele-specific ddPCR quantification of *ONECUT1*

transcription (Fig. 2J) confirmed that the same two lines had a small but significant allele-

specific reduction in *ONECUT1* transcription (Fig. 4F).

To investigate the impact of rs528350911 on TF binding at *ONECUT1-e664kb* following

chromatin immuno-precipitation (ChIP), we developed an allele-specific ddPCR assay targeting

the reference and reference and alternate (T2D associated) nucleotides (Fig. 4G). By utilizing the

heterozygous lines, this allele-specific assay supports an internally controlled experimental

system that mitigates technical variabilities arising from *in vitro* differentiation, ChIP, and PCR.

We selected two hPSC lines that contained the risk variant on different alleles; one line had a

significant impact on *ONECUT1* expression, while the other did not show a significant effect as

mentioned above. After differentiating the lines to the PFG stage, we performed ChIP for

GATA4, GATA6, and FOXA2. The precipitated DNA was then used as the template in the

ddPCR assay, revealing a ~4-fold reduction in GATA4 binding on the allele containing the

disease-associated SNP in both lines (Fig. 4H). We also observed significantly decreased

GATA6 and FOXA2 binding. Thus, despite a mild effect on *ONECUT1* transcription, the disease

SNP substantially reduces TF binding to *ONECUT1e-664kb*.

Discussion

Pancreatic developmental enhancers have been found in mice by integrating pre-existing

knowledge of important pancreatic genes with annotations such as evolutionary conservation and

DNA accessibility. Using this approach, several murine *PDX1* enhancers were identified, and

knockout mice had a spectrum of pancreatic phenotypes depending on the specific region that

was removed[82,83]. A similar approach also identified a putative *GATA4* enhancer, which was

subsequently verified by a transgenic reporter assay in endoderm and endoderm-derived tissues

including the pancreas and duodenum[84]. These studies set the stage for integration of

biochemical epigenetic marks into the enhancer discovery framework, which led to the discovery

of a *PTF1a* enhanceropathy responsible for pancreatic agenesis and monogenic diabetes in

humans[12,13]. The studies also highlight the value of developmental enhancer discovery for

elucidating mechanisms of organogenesis and disease diagnosis. Building upon these findings,

our study expanded the scope of enhancer discovery by targeting developmentally accessible

chromatin regions for large-scale functional interrogation using dCas9-KRAB. Our study not

only identified the human *PDX1* and *GATA4* enhancers corresponding to the aforementioned

mouse enhancers (Sup. Table 1), but also uncovered additional enhancers that had not been

previously described. These newly discovered can serve as a diagnostic resource not only for the

large number of monogenic diabetes patients with no known genetic cause[14] but also for

comprehending variants identified through population-level association studies.

The identification of *ONECUT1e-664kb* adds to a select group of well-documented long-range

regulatory sequences that activate transcription over genomic intervals exceeding 500 kb[85-93].

These enhancers offer a valuable model for further investigation into long-range transcriptional

regulation and its potential implications in human disease. Our findings underscore the value of

unbiased discovery, and we speculate that gene regulation by long-range enhancers is more

common than appreciated. From a technical standpoint, the substantial genomic distance between

ONECUT1e-664kb and its target gene presented a challenge for definitively demonstrating the

cis impact of enhancer deletion. In addition to conducting Hi-C to support 3D enhancer-promoter

interaction, we phased the variants in the *ONECUT1* locus and employed high-sensitivity

variant-specific ddPCR assays in heterozygous lines. As a result, we uncovered two mechanisms

beyond transcriptional control by which regulatory regions could influence disease traits: loss of

promoter antisense transcription and H3K27ac in a genomic locus. We also demonstrate a near-

complete, allele-specific loss of *ONECUT1* transcription resulting from the deletion of

ONECUT1e-664kb. Considering results from *ONECUT1* knockout studies in mice[53] and hPSC

pancreatic differentiation[28], we infer ONECUT1 protein loss to be the primary driver of the

pancreatic differentiation phenotype in our study. Our findings are also in concordance with the

reduced number of PDX1 expressing cells during pancreatic budding in *ONECUT1* null mice

that contributes to pancreatic hypoplasia[53]. We predict that loss of *ONECUT1e-664kb* in an

individual would phenocopy the severe pancreatic hypoplasia observed in cases of *ONECUT1*

protein-coding mutations in clinical settings[28,29].

In previous GWAS, the variant rs528350911 is assigned to the closest gene – *WDR72* [27].

Considering the strong effect of *ONECUT1e-664kb* deletion on *ONECUT1* expression and the

known association of *ONECUT1* with diabetes, this variant most probably confers disease

susceptibility through its effect on ONECUT1 expression. Our work and demonstrates how the

integration of functional screening, precise genome editing, deep learning models, and 3D

genomic information can improve the functional pairing of variants with genes. The introduction

of the disease variant into hPSCs did not have a strong effect on *ONECUT1* transcription, but its

robust effect on TF binding to *ONECUT1e-664kb* suggests that assays other than gene

expression could be valuable for characterizing the molecular consequences of variants in

endogenous loci. We speculate that the rs528350911 risk variant may represent a latent disease

vulnerability, which could be exposed by factors not fully modeled in our current differentiation

system such as age and environment[94,95,]. Given the considerable distance between *ONECUT1e-*

664kb and the *ONECUT1* gene, as well as the limited impact of the rs528350911 risk variant on

ONECUT1 transcription, it is unlikely that this variant would have been prioritized for

investigation in a high-throughput variant modeling screen that relies on prime or base editing

technologies. Higher assay sensitivity is likely necessary to detect the subtle effects of a point

variant[96] like rs528350911 compared to the more pronounced effects of enhancer repression with

dCas9-KRAB[97]. Our work demonstrates that integrating *in vitro* functional enhancer discovery

with population-level genomics provides a rational approach to efficiently prioritize variants for

deeper investigation. This framework holds promise for understanding the exponentially growing

set of disease-associated variants discovered through population-wide GWAS and whole genome

sequencing efforts.

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Competing interests

Authors declare that they have no competing interests.

STAR Methods

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RESOURCE AVAILABILITY

Lead Contact

437 Further information and requests for resources and reagents should be directed to Danwei

438 Huangfu (huangfud@mskcc.org).

Materials availability

440 All the hPSC cell lines and plasmids generated in this study are available upon reasonable

441 request.

Data and Code availability

All the sequencing data generated in this study is available at GEO under accession codes

GSE261390, GSE261391, GSE261392, GSE263174. The Hi-C data from the PFG cells are

available in the 4DN Nucleome Data Portal (https://data.4dnucleome.org) under accession

number 4DNFIQBPRZZD. Previously published data that were re-analyzed here are available at

GEO under code GSE114102, and the 4DN Nucleome Data Portal, accession numbers:

4DNFIKCEWMIA, 4DNFI7RGXYFY, 4DNFIRLG5UWL. Source data are provided with this

paper. All other data supporting the findings of this study are available from the leading author

upon reasonable request. Publicly available software and packages were used throughout this

study according to each developer's instructions.

EXPERIMENTAL MODEL AND STUDY PARTICIPANT DETAILS

Primary cells

The parental hESC line H1 (NIHhESC-10-0043) is available from WiCell under a material transfer agreement.

METHOD DETAILS

hPSC culture

Experiments were performed with H1 human embryonic stem cells (NIHhESC-10-0043), which were regularly confirmed to be mycoplasma-free by the Memorial Sloan Kettering Cancer Center (MSKCC) Antibody & Bioresource Core Facility. All experiments were conducted per NIH guidelines and approved by the Tri-SCI Embryonic Stem Cell Research Oversight (ESCRO) Committee. hPSCs were maintained in Essential 8 (E8) medium (Thermo Fisher Scientific, A1517001) on vitronectin (Thermo Fisher Scientific, A14700) pre-coated plates at 37°C with 5% CO_2. The Rho-associated protein kinase (ROCK) inhibitor Y-276325 (5 µM; Selleck Chemicals, S1049) was added to the E8 medium immediately after passaging or thawing of hPSCs for 24 hours.

hPSC directed pancreatic differentiation

hPSCs were maintained in E8 medium for 2 days to reach ~80% confluence. Cells were washed with PBS and differentiated to DE, GT, and PFG stages following previously described protocols[17,98]. hPSCs were rinsed with PBS and first differentiated into DE using S1/2 medium supplemented with 100 ng/ml activin A (Bon Opus Biosciences) for 3 days and CHIR99021 (Stemgent, 04-0004-10) for 2 days (first day, 5 µM; second day, 0.5 µM). DE cells were rinsed with PBS and then exposed to S1/2 medium supplemented with 50 ng/ml KGF (FGF7)

(PeproTech, 100-19) and 0.25 mM vitamin C (VitC) (Sigma-Aldrich, A4544) for 2 days to reach

GT stage. For PFG stage, cells were switched to S3/4 medium supplemented with 50 ng/ml

FGF7, 0.25 mM VitC and 1 μM retinoic acid (RA) (Sigma-Aldrich, R2625) for 2 days to reach

PFG stage.

gRNA library assembly

23 transcription factors expressed in pancreatic differentiation were selected based on known

clinical roles in metabolic disease and demonstrated pancreatic functional requirement from

CRISPR screens[29,54-57]. Around each transcription factor, loci within hg19 were defined to

include at least the TAD containing the gene, with TADs called on PFG-stage Hi-C data as

previously done[65]. To select putative enhancers for interrogation in these loci, Genrich (0.6.1)[99]

was used to call peaks at an FDR-adjusted p-value of 0.01 on previously published DE, GT, and

PFG stage ATAC-seq data (Sup. Table 1)[25]. For each region in the loci, and some additional

regions, CHOPCHOPv3 was used to design gRNAs for each region[100]. For regions <750 bp, the

top 5 ranked gRNAs were taken, and for regions >750 bp, 1 gRNA/150bp of sequence was

selected for interrogation. Finally, 1100 safe harbor[63], 463 non-targeting gRNAs[64] were included

in the library for a total of 38,747 gRNAs. Oligos were synthesized, amplified, and restriction

cloned into lentiGuide-puro[101] (Addgene; 52963) backbone by the MSKCC Gene Editing &

Screening Core Facility. Cloned plasmid libraries were PCR amplified to incorporate adapters

for NGS. Samples were purified and sequenced using Illumina HiSeq 2500 platform. FASTQ

files were clipped by position and reads were mapped back to the reference library file to show

relative abundance of reads per gRNA. Reads within each sample were normalized to total

number of mapped reads and library size. The overall representation of the libraries was charted

over a one-log fold change to evaluate if any gRNA was over- or under-represented in the final

library.

Lentiviral gRNA libraries

gRNA lentiviral library generation was performed as previously described[17]. In brief, a total of

9.45 μg of the library plasmid combined with 6.75 μg lentiviral packaging vector psPAX2 and

1.36 μg vesicular stomatitis virus G (VSV-G) envelope expressing plasmid pMD2.G (Addgene

plasmids 12260 and 12259) were transfected with the JetPRIME (VMR; 89137972) reagent into

10E6 293T cells in a 10cm plate to produce the lentiviral particles. Fresh medium was changed

24h after transfection and viral supernatant was collected at 48 and 72h after transfection, spun at

1000 rpm for 5 minutes, 0.45um filtered, and stored at −80°C.

CRISPR dCas9-KRAB screens

Two independent screens were performed following the same procedure. H1 hPSC (i)dCas9-

KRAB cells were used[65]. The lentiviral library was transduced into 12E6 hPSCs distributed in 3

x 10 cm plates following the conditions (MOI: 4, protamine sulfate: 6 μg/ml, Ri: 10 μM). After

one day of recovery, doxycycline (DOX) was added and cells were kept under puromycin

selection for 2 days. After two days, 30E6 selected cells were seeded into 6-well plates (5E5 per

well) in VTN-coated plates, grown for 48 hours in E8 medium. After washing with Phosphate

Buffered Saline (PBS) w/o Ca2+ & Mg2+, cells were differentiated to PFG cells, with

doxycycline added throughout the nine-days differentiation process. After, cells were stained

with LIVE/DEAD reagent, fixed, and stained with the PDX1 antibody, and then a fluorescent

secondary antibody. They were then sorted by the MSKCC cytometry facility based on 2ndary

antibody fluorescence using FACSAria sorters (BD Biosciences). Positive and negative cells

were sorted aiming for a minimum 1000X representation (number of cells sorted*MOI)/(unique

gRNAs in library) per condition. For replicate 1, approximately 10E6 PDX1- and 10E6 PDX1+

cells were sorted. For replicate 2, approximately 20E6 PDX1- and 20E6 PDX1+ cells were

sorted. Cells were pelleted and kept at -80°C for downstream gRNA sequencing.

gRNA sequencing and screen analysis

gRNA enrichment sequencing was performed by MSKCC Gene Editing & Screening Core

Facility as previously described, but with modifications to enable DNA extraction from fixed

cells[17]. Cells were decrosslinked and treated with proteinase K (0.1 mg/ml) overnight followed

by genomic DNA extraction with QIAGEN Blood & Cell Culture DNA Maxi Kit (QIAGEN;

13362) and quantified by Qubit (Thermo-Scientific) following the manufacturer's guidelines. A

quantity of gDNA covering the representation of gRNAs described above was amplified with

oligos containing the Illumina adapters and multiplexing barcodes by PCR. Amplicons were

quantified by Qubit and Bioanalyzer (Agilent) and sequenced on the Illumina HiSeq 2500

platform. The gRNA library sequences were used to align the sequencing reads and the counts

for each gRNA were determined with MAGeCK. Screens were analyzed with MAGeCK RRA

(0.5.9.5), and the two screen replicates were analyzed independently (Sup. Table 1). To calculate

a single value for the magnitude of effect, a ratio was taken between the positive and negative

scores and log-transformed. The top 36 hits associated with a PDX1 decrease had a MAGeCK

neg.FDR<0.1 in replicate 2 and caused a PDX1 decrease in both replicates (Sup. Table 1). Hits

overlapped with annotated promoter regions were designated as "Promoter Hits," with the rest

designated as "Non-Promoter Hits." Hits overlapped with annotated promoter regions were

designated as "Promoter Hits," with the rest designated as "Non-Promoter Hits." Promoter

overlap was defined as overlap with an interrogated region ± 1.5kb of Gencode (v18) TSS sites.

For Figure 1F, if there were multiple overlapped regions, the most significant hit was used for

plotting. To link enhancers to the putatively regulated genes, the nearest linearly adjacent

"Promoter Hit" was inferred to be the regulated gene of the "Non-Promoter Hit." To determine

hit and non-hit overlap with H3K27ac ChIP-seq peaks during differentiation, Genrich (0.6.1)[99]

was used to call peaks at an FDR-adjusted p-value of 0.01 on previously published ES, DE, GT,

and PFG stage H3K27ac ChIP-seq data[25]. Conservation scores of regions were determined by

calculating the average of per nucleotide conservation across each region from PhyloP100 per-

nucleotide conservation scores[67].

Generation of *ONECUT1e-664kb* clonal knockout hPSC lines

H iCas9 hPSC lines[60] were used to generate three enhancer deletion lines carrying ONECUT1e-

664kb deletions. Two gRNAs targeting the flanking regions of the enhancer were designed, and

knockouts generated as previously described but with some modifications[8]. gRNAs and tracer

RNA were ordered from IDT (Alt-R CRISPR-Cas9 crRNA and 1072532) and added at a 15 nM

final concentration. In brief, gRNA/tracer RNA and Lipofectamine RNAiMAX (Thermo Fisher

Scientific, 13778030) were diluted separately in Opti-MEM (Invitrogen, 31985070), mixed

together and incubated for 15 min at room temperature, and added dropwise to freshly seeded

iCas9 hPSCs in a 24-well plate. Doxycycline (2 μg/ml) was added the day before transfection,

the day of transfection and 1 day after transfection to induce the Cas9 expression. Three days

after transfection, hPSCs were dissociated into single cells and ~1-3E3 cells were plated into one

100-mm tissue culture dish for colony formation. After ~10 days of expansion, single colonies

were picked. Genomic DNA from crude cell lysate was used for PCR genotyping (primers:

ONECUT1e_KOgenot_F, ONECUT1e_KOgenot_R). Heterozygous lines contained a band

corresponding to the WT band and a smaller band corresponding the knockout band, while

homozygous lines only had smaller bands. Further genotyping was done through TOPO cloning

of the genotyping amplicons, which determined the exact genotypes of both alleles for all lines

(Sup. Table 2). gRNA target sequences and primers used for PCR and sequencing are listed in

Supplementary Table 3.

Generation of clonal hPSC lines with disease-associated variant

A process similar to enhancer deletion line generation was used to generate heterozygous knock-

in SNP lines carrying the rs528350911 variant. During transfection, a single gRNA proximal to

the variant site was used, and an ssDNA HDR template containing the disease-associated variant

was added to the mixture at the Lipofectamine transfection step (Sup. Table 3) at a concentration

of 20 nM. Genomic DNA from crude cell lysate was used for PCR (primers:

ONECUT1e_KOgenot_F, ONECUT1e_KOgenot_R). All picked clones were sequenced, and

heterozygous lines picked by the presence of a double peak on the spectrogram that

corresponded to the insertion of the disease-associated variant. For lines containing the

introduced variant, no indels or other changes from the reference genome were detected in the

amplicon.

ONECUT1 locus SNP phasing

To phase all SNPs in the *ONECUT1* locus, Nanopore sequencing with adaptive sampling of the

region chr15:52,707,645-53,586,740 (and controls regions) was conducted on an H1 iCas9 line

with the rs528350911 heterozygous variant (line N11). DNA was extracted using Monarch®

HMW DNA Extraction Kit for Cells & Blood (NEB, T3050), according to the manufacturer's

instructions. DNA quantity was measured using a QuantiT (Thermo Fisher Scientific), purity on

a NanoDrop (Thermo Fisher Scientific) and fragment-size distribution on a TapeStation

(Agilent). Prior to ONT library preparations, DNA was sheared to ~15-20 kb fragment size using

Covaris g-TUBE. Sequencing libraries were prepared from ~1–2ug of DNA, using native library

preparation kit, SQK-LSK114, according to the manufacturer's instructions. Each library was

loaded onto a PromethION flow cell, R10.4.1, and sequenced on an ONT PromethION P24

device. The sample was run for 72 h, with 1 nuclease flush and reload performed during the run

to maximize sequencing yield.

Reads were aligned with minimap2 (2.26)[102], with samtools[103] coverage estimating ~138X

coverage across the region enriched with adaptive sampling. WhatsHap (2.2)[104] was used to

phase the variants in the locus as well as assemble a haplotype containing the *ONECUT1* gene,

enhancer, and surrounding regions (Sup. Fig. 2D). WhatsHap was also used to phase-tag reads

for visualization in Sup. Fig. D and E. From these results, a heterozygous SNP was identified in

the 3'UTR of the *ONECUT1* transcript, as well as the enhancer-proximal Genotyping SNP.

ONECUT1e-664kb heterozygous knockout line phasing

To assess which allele contained the knocked-out enhancer in the heterozygous enhancer

deletion lines (Het. 1, Het. 2), a forward primer was designed centromeric to the Genotyping

SNP and a reverse primer targeting the enhancer sequence that would be present on the non-

deletion allele. Since line designated as Het. 1 contained an inversion of the WT enhancer

sequence, two different reverse primers were needed. Thus, for Het. 1, and amplicon was

generated with the primer pair ONECUT1e_GenotSNP_F,

ONECUT1e_GenotSNP_R_WT_inv_e, and for Het.2, the pair ONECUT1e_GenotSNP_F,

ONECUT1e_GenotSNP_R_WT_e. The amplicon was sent for sanger sequencing with the

ONECUT1e_GenotSNP_F primer used as the sequencing primer. The Genotyping SNP resulting

from this sequencing was used to determine the allele containing the deletion.

ONECUT1e-664kb heterozygous rs528350911 knock-in line phasing

Amplicons were generated with the genotyping primers (ONECUT1e_KOgenot_F,

ONECUT1e_KOgenot_R) and sanger sequenced with both primers. To phase the variant, we

conducted a PCR amplifying an ~8.1kb fragment containing both the T2D-associated variant and

the Genotyping SNP (primers: ONECUT1e_GenotSNP_F , ONECUT1e_GenotSNP_R_WT_e).

The amplicon was sent to Plasmidsaurus for linear fragment sequencing. Reads were aligned

with minimap2 (2.26), and variants were phased with longshot (0.4.5)[105].

Immunofluorescence staining

In vitro pancreatic differentiated cells were fixed in 4% paraformaldehyde (Thermo Fisher

Scientific, 50980495) for 10 min at room temperature. After washing with PBST (PBS with

0.1% Triton X-100) three times, cells were blocked in 5% donkey serum in PBST buffer for

30 min at room temperature. Primary and second antibodies were diluted in the blocking solution

(Sup. Table 4). Cells were incubated with primary antibodies overnight at 4°C, followed by 1 hr

staining for secondary antibodies at room temperature. The cells were stained with 4′,6-

diamidino-2-phenylindole (DAPI) for ~15 min at room temperature. Images were taken using a

confocal laser scanning platform (Leica TCS SP5).

Flow cytometry analysis

Cells were dissociated using TrypLE Select and resuspended in FACS buffer (5% FBS in PBS).

Live-Dead Fixable Violet Dead cell stain (Invitrogen, L34955) was used to discriminate dead

cells from live cells. Live/dead staining was performed per manufacturer's instructions in FACS

buffer, followed by fixation and intracellular staining with a FOXP3 staining buffer set

(eBioscience, 00-5523-00) following the manufacturer's instructions. Permeabilization/fixation

was performed at room temperature for 1 hr. Antibody staining was performed in

permeabilization buffer. Antibodies for this study are listed in Supplementary Table 8. Cells

were then analyzed using BD LSRFortessa. Flow cytometry analysis and figures were generated using FlowJo v.10.

RNA-seq and analysis

RNA was extracted from differentiated cells with the Zymo Quick-RNA MiniPrep kit (R1055). TruSeq stranded total (Ribodepletion) RNA-seq was performed on three independent differentiations by the MSKCC Integrated Genomics Operation (IGO) Core. Briefly, STAR alignment was performed followed by featureCounts[106] quantification (Sup. Table 6) and normalization with DESeq2 median of ratios prior to visualization in Fig. 3A, Fig. 3C, and Sup. Fig. 3B. Differential gene expression was performed and adjusted p-values computed with DESeq2. Strand-specific analysis of promoter antisense transcription analysis was performed with featureCounts strand specific quantification of a region telomeric to *ONECUT1* (chr15:52805588-53391232). For promoter antisense transcript visualization, samtools was used to extract sense and antisense alignments from .bam files from which bigwig coverage files were generated.

Droplet digital PCR (ddPCR) and analysis

Assays specific for the detection of SNPs (Sup. Table 5) were designed (using Primer3Plus, if needed) and ordered through Bio-Rad. Cycling conditions were tested to ensure optimal annealing/extension temperature as well as optimal separation of positive from empty droplets. Optimization was done with a known positive control. A constant amount of RNA or DNA was used within each experimental reaction. Reactions were partitioned into a median of ~17,500 droplets per well using the QX200 droplet generator. Plates were read and analyzed with the QuantaSoft software to assess the number of positive droplets positive.

Procedure for RNA: after PicoGreen quantification of RNA, droplet generation was performed

on a QX200 ddPCR system (Bio-Rad catalog # 1864001) using cDNA generated from 0.1-2 ng

total RNA with the One-Step RT-ddPCR Advanced Kit for Probes (Bio-Rad catalog # 1864021)

according to the manufacturer's protocol with reverse transcription at 42°C and

annealing/extension at 60°C.

Procedure for DNA: after PicoGreen quantification of DNA, 0.25-9 ng gDNA or ChIP-ed DNA

was combined with locus-specific primers; FAM- and HEX-labeled probes; Hae III (DNA

extracted form hPSC SNP lines), Hind III (ChIP-ed DNA), or Mse I (DNA extracted

ONECUT1e-664kb hPSC knockout lines); and digital PCR Supermix for probes (no dUTP).

Emulsified PCRs were run on a 96-well thermal cycler using cycling conditions identified during

the optimization step (95°C 10'; 40 cycles of 94°C 30' and 53°C (rs528350911 assays only) or

60°C 1'; 98°C 10'; 4°C hold).

Hi-C and analysis

Hi-C was performed as previously described with cells differentiated to the PFG stage[8]. HiC-Pro

(3.1.0)[107] was used to align the individual Hi-C replicates to GRCh38, GCA_000001405.15,

with alternative contigs removed, and then final merged libraries were obtained by combining all

*.allValidPairs files where duplicates across biological replicates are maintained. DE libraries

consisted of three biological replicates (two HUES8, one H1 with accessions

4DNFIKCEWMIA.hic, 4DNFI7RGXYFY.hic, and 4DNFIRLG5UWL.hic respectively), while

the PFG libraries consisted of two biological replicates differentiated to the PFG stage (merged

accession 4DNFIQBPRZZD.hic). For visualization, Hi-C data was visualized at 25 KB

resolution.

Virtual 4C counts for the regions of interest were extracted using utilities from plotGardener

(1.6.4)[108] and plotted as a one-dimensional signal. Hi-C matrix counts for the regions of interest

were also extracted using plotGardener and plotted as a rectangular region. Observed counts are

library normalized prior to plotting so that the total number of reads between panels is

comparable. TADs that were previously identified were also overlaid on the data. Counts within

the vicinity of the *ONECUT1* promoter, all contacts whose start or end bins sit within a 1.2Mb

window around the *ONECUT1* promoter were extracted with strawr (0.0.91) and passed to

DESeq2 (1.40.1)[109]. The Wald test was used to assess significance of all differential contacts

within that window, and significant hits anchored on the *ONECUT1* containing bin are reported

at an FDR of 0.1.

For calling topologically associating domains, merged Hi-C files were first ICE normalized

through HiTC (1.34.0)[110] and then topologically associating domains were identified with Hi-

CDC+ (1.5.2)[111] using its interface to TopDom at 50 kb. Window size for scanning was set at 5.

ChIP and ChIP-seq assays

ChIP was performed as previously described[18]. Briefly, for each sample, around 30E6 cells were

crosslinked with 1% formaldehyde for 10 minutes at 37°C and quenched with 0.125 M glycine.

Fixed cells were collected in cold PBS buffer and washed with the same. Then the cells were

snap frozen and saved at -80°C. On the day of immunoprecipitation, cell pellet was thawed on

ice, resuspended in 700uL SDS buffer (1% SDS, 10 mM EDTA, 50 mM Tris–HCl, pH 8), and

incubated for 10min on ice. Sonication was performed on a Branson Sonifier 150 set at 30%

amplitude for 5.5min (10s on/off pulsing). Supernatant was pre-cleared with Dynabeads Protein

G (#10004D) and then incubated overnight with antibodies (10 ug for transcription factors, 5 ug

for H3K27ac, Sup. Table 4) at 4°C.). On the next day, the ChIP samples were incubated with

Dynabeads Protein G for 6 hours at 4°C. Then the beads were pelleted and sequentially washed

twice with low salt (0.1% SDS, 1% Triton X-100, 2 mM EDTA, 20 mM Tris–HCl, pH 8, 150

mM NaCl), then high salt (0.1% SDS, 1% Triton X-100, 2 mM EDTA, 20 mM Tris–HCl, pH 8,

500 mM NaCl), and finally TE buffer (10 mM Tris–HCl, pH 8, 1 mM EDTA). The DNA was

eluted from the beads by incubating in elution buffer (1% SDS, 0.1 M NaHCO3) at 65°C for

15min and decrosslinked with 190 mM NaCl at 65°C overnight. A total of 5 μl 0.5 M EDTA, 10

μl 1 M Tris–HCl, pH 6.5, and 1 μl Proteinase K (20 mg/ml) were added to 260 uL of the de-

crosslinked product and incubated for 1 hr at 45°C. DNA was isolated by using QIAquick PCR

purification kit (Qiagen, 28104; QIAGEN). H3K27ac ChIP samples were submitted to MSKCC

Integrated Genomics Operation core for NGS library preparation and sequencing. GATA4,

GATA6, and FOXA2 ChIP samples were submitted for ddPCR.

H3K27ac ChIP-seq Analysis

Sequenced data were aligned to the hg38 reference genome using bowtie2 (2.5.1)[112]. Peak calling

was performed using MACS2 (2.2.9.1)[113] with the respective ChIP input as the control and the

default P value cutoff and default extension size. Irreproducible discovery rate (IDR)[114] with a

cutoff of 0.01 was used to filter pick up reproducible peaks, where at least 2 pairwise

comparisons of the three replicates passed the cutoff. Peaks were filtered by ENCODE exclusion

list version 2 for hg38[115]. The quantification and differential signal intensity of the samples were

calculated in DESeq2[109]. The integer counts, log(fold-change), and adjusted P values are

provided in Supplementary table 7). The replicates were combined and the signal track was

generated from MACS2. We used MACS2 subtract to generate the difference of the signal track.

All signal tracks were visualized using Integrative Genomics Viewer (IGV)[116].

For allele-specific analysis, GATK ASEReadCounter was used to quantify reads at heterozygous

positions phased by long-read sequencing and ratios were calculated for heterozygous positions

(score filters: PHASE_QUAL >= 140, and QUAL >= 200) with more than 100 reads total across

all WT and Het. 2 replicates. For figure 4F, the "Phased Region" is the haplotype block

containing *ONECUT1* called by WhatsHap (chr15:50462878-54550826) and indicated in Sup.

Fig. 2D.

chromBPNet variant modeling

A ChromBPNet[78,79] model was trained to predict the ATAC-seq profile using bulk ATAC-seq

data from the PFG stage[25]. We utilized default hyperparameters with an input size of 2114 base

pairs and an output size of 1000 base pairs for training both the bias model and the ChromBPNet

model. For the bias model training, we selected regions that do not overlap with any ATAC-seq

peaks in the four stages of pancreas differentiation[25], applying a bias threshold factor of 0.3 to

filter out high-count regions in the sets. This resulted in a training set of 54957 regions. For

ChromBPNet training, we used ATAC-seq peaks only from the PFG stage with IDR values

greater than 830, resulting in a training set of 52541 regions. During training for both models, we

held out peaks from chromosomes 15 and 13 for validation. For each specific region of interest,

we used the 2114 base pairs around the summit center as input and obtained the DeepLift

attribution score using the trained ChromBPNet model. Specifically, we employed the original

attribution pipeline from the ChromBPNet GitHub repository, which involves computing the

DeepLift score over 20 reference inputs shuffled from the original sequence. To find the GATA

specific motifs, we ran FIMO[117] on our region of interest using the positional weight matrix

(PWMs) of all GATA variants from the CIS-BP2.00 dataset[118], and filtered out the motif hits

with p-values < 0.05.

QUANTIFICATION AND STATISTICAL ANALYSIS

All datapoints refer to biological repeats. No statistical method was used to predetermine sample

sizes. The investigators were not blinded to allocation during experiments and outcome

assessment. No data were excluded from the analyses unless the differentiation experiment itself

failed. The number of biological and technical replicates are reported in the legend of each

figure. Flow cytometry analysis were derived from at least three independent experiments unless

specified in the legends. For ChIP-seq and bulk RNA-seq, quantification and statistics were

derived from at least three independent experiments. CRISPR dCas9-KRAB screening was

performed twice. Quantification is shown as the mean $\pm$ s.d. All the statistical analysis methods

are indicated in the figure legends and methods.

Fig. 1

Fig. 1: CRISPRi repression screen to discover pancreatic differentiation enhancers

A) hPSC stepwise pancreatic differentiation protocol schematic. ActA, activin A; CHIR, CHIR99021; RA, retinoic acid; VitC, vitamin C. B) Gene selection rationale for enhancer discovery screen.

C) Putative enhancer region selection and gRNA design schematic.

D) Screening procedure schematic.

E) Results from two screen replicates, each point represents a genomic region. 36 regions with a FDR < 0.1 in replicate 2 and have a decrease in PDX1 in both replicates.

F) Results from screen replicate 2 compared to whole genome PDX1 expression screen. Genes showing enrichment in both screens are labeled.

G) Region overlap with H3K27ac peaks at any stage during differentiation (ES, DE, GT, PFG).

H) Differentiation stages at which non-promoter screen hits overlap with H3K27ac peaks, and the number of hits with each activation pattern.

I) Enhancer-gene pair assignment.

J) Enhancer-gene pair linear distances.

Fig. 2

Fig. 2: *ONECUT1e-664kb* deletion and pancreatic differentiation characterization

A) Hi-C data at the definitive endoderm and posterior foregut stage differentiated hPSCs are shown around the *ONECUT1* locus. Contact between *ONECUT1e-664kb* and *ONECUT1* promoter region are circled to show the increase in contact frequency between the two loci. Red lines demarcate TAD boundaries identified at 50 kb.

B) 1D slices of the 2D contact map were obtained with the *ONECUT1* promoter set as the bin of origin. Hi-C data was library scaled and shows the changes observed in (A). Significantly different bins were determined by DESeq2.

C) H1 hPSC enhancer deletion line schematic and genotyping PCR.

D) Representative flow cytometry plots of WT and enhancer deletion cells differentiated to the PFG stage.

E) Proportion of cells achieving a ONECUT1+ identity. Each dot represents one independent experiment (n = 5 independent experiments) and data are presented as the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

F) Proportion of cells achieving a PDX1+ identity. Each dot represents one independent experiment (n = 5 independent experiments) and data are presented as the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

G) Representative histograms of ONECUT1 MFI from flow cytometry plots of unedited and enhancer knockout cells differentiated to the PFG stage. Quantification of all replicates of ONECUT1 MFI of PDX1+ cells.

H) Schematic for identifying ddPCR and Genotyping SNPs.

I) Spectrogram of genotyping SNP sequencing and schematic demonstrating allele that contained the deletion in heterozygous lines.

J) ddPCR assay design schematic.

K) Allele-specific *ONECUT1* transcript quantification with ddPCR on RNA extracted from PFG stage cells. Each allele is shown in a separate plot, with the expression ratio shown in the last plot. Each symbol represents one independent differentiation (n = 3 independent experiments), average of two technical replicates per differentiation. Data are presented as the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

Fig. 3

Fig. 3: Transcriptional and epigenetic consequences of *ONECUT1e-664kb* knockout

A) RNA-seq normalized read counts for genes in *ONECUT1* locus at GT and PFG stages from WT, heterozygous (het. 2), and homozygous enhancer knockout cells. * indicates padj<0.005 computed by DESeq2.

B) Visualization of *ONECUT1* promoter antisense RNA read coverage from strand-specific RNA-seq of PFG-stage WT and homozygous enhancer knockout cells (representative differentiation replicate).

C) Quantification of (B) from three differentiations. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

D) H3K27ac ChIP-seq at PFG stage of WT, heterozygous, and homozygous enhancer knockout cells. TADs called from PFG stage Hi-C data. Difference calculated via MACS2 subtract. Differential peaks between WT and homozygous enhancer knockout cells quantified with DESeq2. *ONECUT1e-664kb* deletion region is within peak #12.

E) H3K27ac is decreased in an allele-specific manner within significantly affected H3K27ac peaks.

F) Quantification of average H3K27ac read ratios at all heterozygous SNP positions in phased locus in top significantly different peaks (WT vs Homo, padj<1E-15) and other peaks. Each point represents a heterozygous SNP position, average of three replicates. Ordinary one-way ANOVA with Šidák's multiple comparisons test, with a single pooled variance.

835   **Fig. 4:**

Fig. 4: T2D-associated SNP hPSC modeling and pancreatic differentiation characterization

A) LocusZoom plot of variants from Mahajan et al (2018b) T2D GWAS meta-analysis unadjusted for BMI. Variants within credible sets are colored by the PPA score.

B) Decreased DeepLift attribution score at GATA4 motif upon introduction of variant. Red box represents a GATA4/6 binding motif hit by FIMO (p < 0.001).

C) Spectrogram of disease-associated SNP sequencing and schematic demonstrating allele that contains the introduced SNP.

D) Proportion of ONECUT1+ cells at PFG stage. Each symbol represents one independent experiment (n = 6 independent experiments) and data are presented as the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

E) ONECUT1 MFI of PDX1+ cells. Each symbol represents one independent experiment (n = 6 independent experiments) and data are presented as the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

F) *ONECUT1* transcript ratio of edited allele divided by unedited allele. Each symbol represents one independent differentiation (n = 6 independent experiments), average of two technical replicates per differentiation. Data are presented as the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

G) ChIP-ddPCR schematic.

H) ChIPed DNA ratio.Each symbol represents one independent differentiation and ChIP experiment, average of two technical replicates per differentiation. Two-way analysis of variance (ANOVA) followed by Tukey's multiple comparisons test versus WT control.

Supplemental Information

Supplementary Figures:

Sup. Fig. 1

Sup. Fig. 1, Related to Figure 1: Supplemental results and analysis for CRISPRi screens

A) Representative flow cytometry plot pre- and post-sort PDX1 sorting.

B) Rank plot of replicate 2 MAGeCK scores with FDR < 0.1

C) Calculated conservation scores for each region within indicated group. Ordinary one-

way ANOVA with Šidák's multiple comparisons test, with a single pooled variance.

D) Representative IGV plots for enhancer-gene pair rationale. Each bar represents a

MAGeCK score from replicate 2. A negative bar indicates a region caused net decrease

in the proportion of PDX1+ cells, with a larger bar indicating a more significant change

in PDX1 enrichment. Hollow black points represent promoter region hits, filled black

points represent non-promoter region hits.

E) H3K27ac ChIP-seq and ATAC-seq at representative enhancer loci.

    Sup. Fig. 2

Sup. Fig. 2, Related to Figure 2: *ONECUT1e-664kb* deletion supplemental results and analysis

A) *ONECUT1* transcript ddPCR on RNA from WT cells at GT and PFG stages.

B) PFG-stage immunofluorescence for PDX1 and ONECUT1 of WT and homozygous enhancer knockout cells.

C) *ONECUT1* transcript ddPCR on DNA from WT and enhancer knockout cells. One-way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test versus WT control.

D) Plot of phased reads and haplotype blocks from nanopore sequencing of H1 hPSCs with adaptive sampling.

    Sup. Fig. 3, Related to Figure 3: Supplemental RNA-seq and H3K27ac ChIP-seq results
    and analysis

        A) Non allele-specific *WDR72* ddPCR results of WT and enhancer knockout cells
        differentiated to the PFG stage.
        B) Visualization of sense read coverage in region telomeric to *ONECUT1* from strand-
        specific RNA-seq of PFG-stage WT and homozygous enhancer knockout cells
        (representative differentiation replicate). Quantification from three differentiations. One-
        way analysis of variance (ANOVA) followed by Dunnett multiple comparisons test
        versus WT control.
        C) Full chromosome view of H3K27ac in WT and enhancer knockout cells showing the
        ONECUT1 TAD has the most significantly affected regions, as well as the regions with
        the greatest magnitude loss of H3K27ac.
        D) Allele-specific H3K27ac ChIP-seq analysis schematic.

Sup. Fig. 4

Sup. Fig. 4, Related to Figure 4: Supplemental results for T2D-associated SNP hPSC
modeling
A) Visualization of ChIP-seq at *ONECUT1-e664kb* locus conducted at PFG stage. Black
bar indicates *ONECUT1e-664kb*.
B) Bar plots of proportion of cells achieving a PDX1+ identity. Each symbol represents
one independent experiment (n = 6 independent experiments) and data are presented as
the mean ± s.d. One-way analysis of variance (ANOVA) followed by Dunnett multiple
comparisons test versus WT control.
918

Supplementary Tables:

Table S1: Enhancer screen region selection and analysis results. Sheet 1: Transcription factor loci selected for CRISPRi interrogation. Sheet 2: ATAC-seq accessible regions selected for interrogation. Sheet 3: Read abundance quantification of each gRNA. Sheet 4: MAGeCK-RRA region-level analysis of two screens. Sheet 5: Top screen hits, FDR<0.1 in screen 2.

Table S2: Engineered hPSC line allele genotypes.

Table S3: List of gRNA target sequences and primer sequences for PCR genotyping.

Table S4: Antibody target, company or lab that supplied antibody, catalog number, application, and quantity or dilution used are indicated. For ChIP experiments, the amount used for cells from around half of a 15cm plate (roughly 40-50 million cells) is indicated.

Table S5: ddPCR assay details.

Table S6: RNA-seq conducted on WT, heterozygous (het. 2) and homozygous *ONECUT1e-664kb* enhancer knockout cells. Sheet 1: DESeq2 normalized gene counts. Sheet 2: DESeq2 differential analysis between WT and heterozygous enhancer knockout cells at PFG stage. Sheet 3: DESeq2 differential analysis between WT and homozygous enhancer knockout cells at PFG stage.

Table S7: H3K27ac ChIP-seq conducted on PFG-stage WT, heterozygous (het. 2), and homozygous *ONECUT1e-664kb* enhancer knockout cells. Sheet 1: Raw integer counts of H3K27ac reads. Sheet 2: DESeq2 calculated differential statistics between pairwise comparisons among WT, heterozygous, and homozygous enhancer knockout cells. Sheet 3: Phased read counts at indicated position, read ratios for each replicate, average read ratios for each genotype. Top_Differential_Peak indicates whether the position falls within a top differential peak (WT vs Homo, padj>1E-15).

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