

# Supplementary Note:

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## Supplementary Methods

### Genetic Risk Scores for ISI and IFC

We additionally constructed genetic risk scores (GRS) for IFC and ISI including lead variants at primary association signals in Europeans. GRS were constructed using the sum of allele scores for each individual where allele counts were weighted by the effect of this allele on IFC or ISI, adjusted or unadjusted for BMI as appropriate. Weights used in the construction of the GRS are outlined in **Supplementary Table 12**. Allele counts were discrete for genotyped variants (0 = reference, 1 = heterozygous, 2 = homozygous) and weights were aligned to the dosage coding allele prior to weighting of each variant. Weighted allele counts were summed for each individual to generate the GRS. GRS were constructed for ISI and IFC adjusted and unadjusted for BMI in unrelated individuals of European ancestry in UK Biobank<sup>1</sup> (N max = 351,987) and the OMICS subset Fenland Study<sup>2</sup> (N max = 8,925). Blood biochemistry measures in the Fenland Study and UK Biobank were natural log transformed prior to scaling. Phenotypes were prepared as has been previously described.<sup>3-5</sup>

Linear regression models were run for each genetic score and phenotype pair adjusting for age, sex and the 1<sup>st</sup> 10 genetic principal components to account for population structure. Phenotypes were scaled to a mean of 0 and SD of 1 to allow direct comparison across phenotypes. Significance was considered at a Bonferroni significance threshold in UKBB ( $P < 4.5 \times 10^{-4}$ ; N GRS = 4, N phenotypes = 28) and nominal significance threshold in the Fenland Study ( $P < 0.05$ ).

### LD Score Regression

We applied LD score regression<sup>6</sup> using python-based software LDSC (accessed at <https://github.com/bulik/ldsc>) to assess genome-wide genetic correlations between IFC and ISI with other cardiometabolic traits. We calculated LD scores for ISI, IFC using BMI adjusted summary statistics from our analyses in cohorts of European ancestry. We further calculated LD scores for additional T2D related traits from publicly available summary statistics for 2 h glucose<sup>7</sup>, fasting insulin<sup>7</sup>, T2D<sup>8</sup> and WHRadjBMI<sup>9</sup> in cohorts of European ancestry. Further pre-computed LD scores were obtained online for additional cardiometabolic traits of interest (<https://data.broadinstitute.org/alkesgroup/LDSCORE/>). The genetic correlations were



calculated restricting to HapMap3 SNPs with a MAF > 0.05, and we used the 1000G European LD Score reference<sup>6</sup>.

### **Tissue specificity**

To dissect the tissue of action of IFC and ISI loci, we used LDSC-SEG to assess the genome-wide enrichment of tissue-specific and cell type specific annotations<sup>10</sup>. In brief, LDSC-SEG employs stratified LD score regression to assess for genome-wide enrichment of sets of specifically expressed autosomal genes for a given trait of interest. We utilised LD scores available for a total of 205 specific tissue and cell types from GTEx (53 tissues; RNA-seq) and Franke Lab dataset which consists of expression data for 152 tissues of cell types in human, mouse, and rat<sup>10</sup>. In addition, we examined LD scores also generated by Finucane *et al* (2018)<sup>10</sup> from chromatin annotation from Roadmap<sup>11</sup> and ENTEEx data from ENCODE<sup>12</sup>. Suggestive enrichment was considered at a nominal significance threshold of  $P < 0.05$ , with no result surviving multiple testing correction.

### **Confirmation of gene knockdown in 3T3-L1 adipocytes**

#### *RNA extraction*

RNA was extracted from 3T3-L1 adipocytes cultured in 24-well plates 4 d after siRNA knockdown (see above). Cells were washed 2x in PBS and TRI reagent (Zymo Research) was added to each well. Cells were then flash frozen in the plate on dry ice and stored at  $-80^{\circ}\text{C}$ . RNA extraction was conducted in an RNase free environment using the Zymo Research Directzol RNA Extraction Kit following standard manufacture protocol. Briefly, plates were thawed on ice and cells were lysed by manual cell scraping and incubation at room temperature for 10 min. Samples were mixed with an equal volume of 100% ethanol and added directly to an RNA extraction column. RNA was eluted in 30  $\mu\text{L}$  of RNase and DNase free water. All centrifugation steps were performed at room temperature at 12,000 x g.

#### *Reverse transcription*

RNA was diluted to a consistent concentration per biological replicate. Random primers (Promega) were annealed in a total volume of 30  $\mu\text{L}$  at  $70^{\circ}\text{C}$  for 5 min, and immediately rapidly cooled on ice, in line with manufacturers' specifications. Reverse transcription was conducted in a total volume of 50  $\mu\text{L}$  for each sample using MMLV Reverse Transcriptase (Promega) and

RNassin RNA inhibitor (Promega). Reverse transcription was conducted using BioRad Tetrad 2 Thermal Cycler at 25°C for 10 minutes, 37°C for 1 h and 85°C for 10 min before being stored at -20°C.

#### *qPCR RNA quantification*

Quantitative PCR was conducted using Thermo Fisher TaqMan gene expression assays following the standard manufacturer protocol. Probe specifications for targets of interest are outlined in **Supplementary Table 31**. These were added to master mix containing TaqMan 2x Universal PCR Master Mix (ThermoFisher) and nuclease-free water. Polymerase chain reaction was carried out using QuantStudio 7 or Quantstudio 5 Real-Time PCR system (Thermo Fisher Scientific). *ActB* (Mm04394036\_g1) and *Canx* (Mm00500330\_m1) were used as housekeeping controls for all samples. The standard pre-set cycling conditions for TaqMan reagents were used.

#### *qPCR data analysis*

Double delta CT values (ddCT) were calculated and the fold change relative to the NT control was used to assess knockdown efficiency. The geometric mean CT value of the housekeeping genes *ActB* and *Canx* were used for normalisation for a given sample. One-way ANOVA was used to compare delta CT (dCT) values for a given knockdown target compared to NT control, and significance was considered at FDR < 5%.

#### *SDS-PAGE and immunoblotting*

Typically, 10 µg of protein was resolved by SDS-PAGE and transferred to PVDF or nitrocellulose membranes. Membranes were blocked in 5% BSA or skim milk powder in TBS-T (0.1% v/v Tween-20 in Tris-buffered saline) for 1 h followed by overnight incubation at 4°C with specific primary antibodies (listed below). Following incubation with horseradish peroxidase (HRP)-conjugated anti-rabbit or mouse immunoglobulin G (IgG) or Alexa-647-conjugated anti-mouse IgG secondary antibodies for 1 h. Protein bands were visualised using ECL (Thermo Scientific) or 647-fluorescence intensity on the Chemidoc MP (Bio-Rad). The following antibodies were used: Lnpep/IRAP (Cell Signaling Technologies, #6918); TBC1D4 (Cell Signaling Technologies, #2670); IRS1 (Cell Signaling Technologies, #3407); PanAkt (Cell Signaling Technologies, #2920); GLUT4 (polyclonal antibody recognising the C-

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## 1. Correlation of post-challenge insulin traits with other metabolic traits in the Fenland Study and RISC cohort

To determine the traits correlated with insulin resistance in the fasted and fed state, we conducted observational correlation analyses for fasting insulin and dynamic insulin traits Modified Stumvoll Insulin Sensitivity Index (ISI) and Insulin Fold Change during an OGTT (IFC), with metabolic traits measured in the Fenland Study (pairwise N= 1454 (c-peptide)-11787). Fasting insulin was positively correlated with many metabolic traits including glycaemic traits fasting and 2hr glucose, proinsulin, HbA1C. Further fasting insulin is positively correlated with LDL triglycerides and liver biomarkers as well as BMI, WHR and blood pressure (**Extended Data Figure 2**).

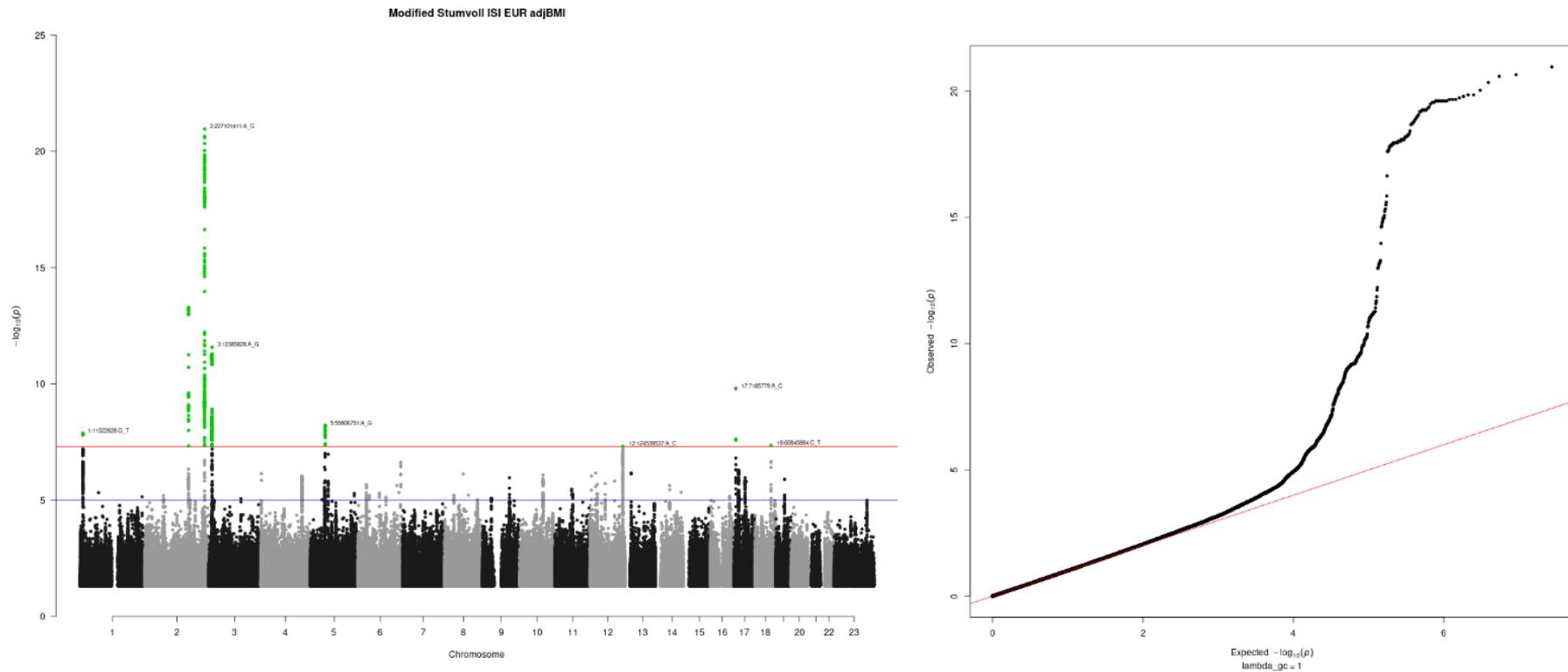
We observed that both IFC and ISI show strong correlation with measures of insulin and glucose taken at 2hrs during a standard OGTT, however are poorly correlated with fasting measures, suggesting these are primarily capturing the post-challenge insulin response. ISI is additionally inversely correlated with liver-related traits, as seen with fasting insulin (hepatic steatosis score and measures of circulating alkaline phosphatase, gamma-glutamyl transferase, alanine aminotransferase), suggesting that ISI may additionally capture insulin action on the liver in the fasted state (**Extended Data Figure 2**). IFC was not correlated with any liver-related traits, suggesting that this potentially more specific to the post-prandial state. This is further evidenced by observations that ISI is more strongly correlated with HOMA-IR, an index of fasting insulin resistance (**Extended Data Figure 2**). Further, IFC was weakly correlated with ISI ( $\rho = -0.35$ ), compared to other insulin related traits e.g. 2 hour insulin, suggesting that a GWAS of this measure may capture variants that are distinct from those that are driving 2-hour insulin or alternatively that these traits are able to capture different aspects and pathways that contribute to the post-challenge insulin response.

We conducted similar analyses in participants without diabetes in the RISC cohort (N=1,138; methods) where hyperinsulinemic-euglycemic clamp and OGTT-derived measures of insulin sensitivity are available. We assessed and compared the correlation of clamp and OGTT-derived measures of insulin sensitivity with relevant traits, as was done in the Fenland Study above. We replicated the strong positive correlation of IFC with insulin at 120 min during an OGTT ( $\rho = 0.725$ ,  $P = 4.21e-177$ ) and saw a weak negative correlation of IFC with fasting

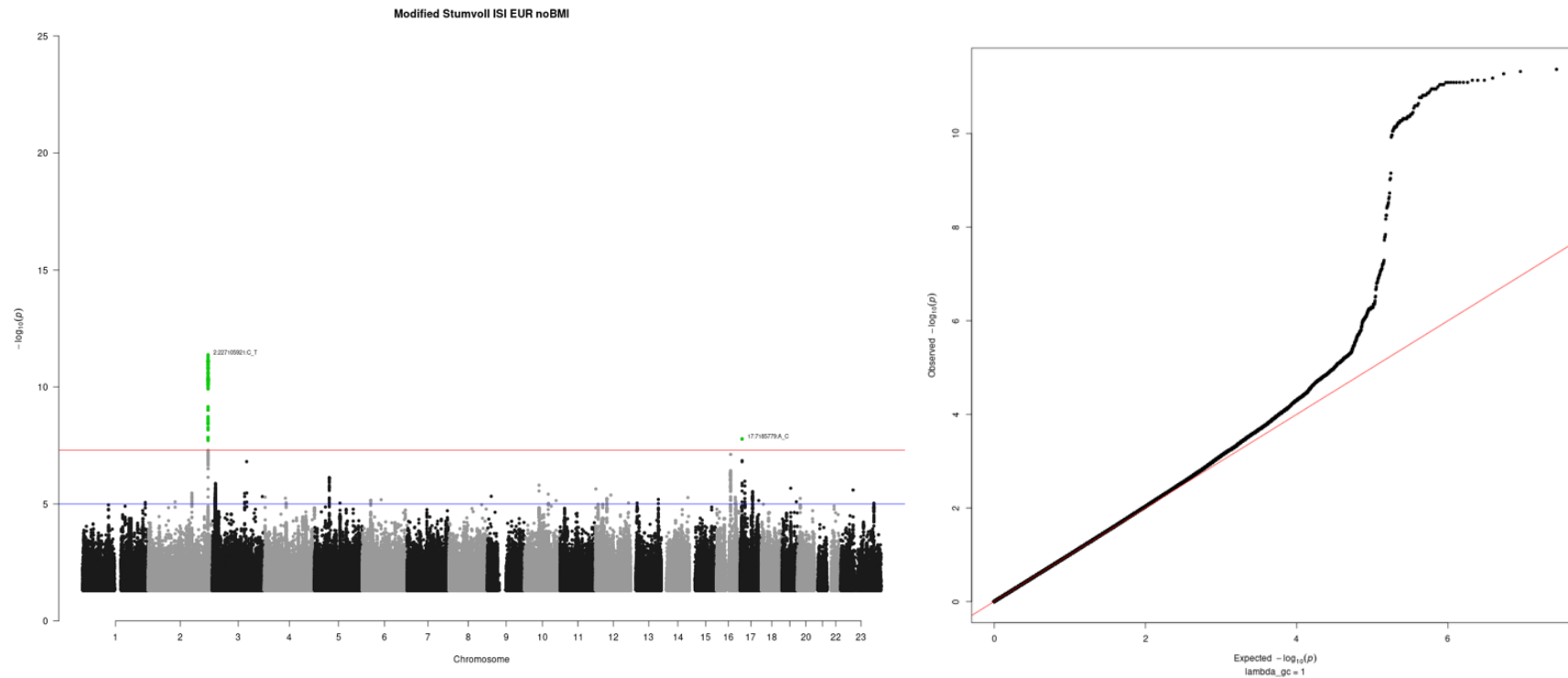
insulin during an OGTT ( $\rho = -0.165$ ,  $P = 4.57e-8$ ) as well as no correlation with insulin taken 20min before clamp ( $\rho = 0.009$ ,  $P = 0.768$ ). We additionally identified a significant inverse correlation between IFC and clamp-derived measure of insulin sensitivity M/I ( $\rho = -0.180$ ,  $P = 3.29e-9$ ). For ISI we identify strong inverse correlation with insulin at fasting ( $\rho = -0.68$ ,  $P = 6.38e-149$ ) and 120mins during an OGTT ( $\rho = -0.869$ ,  $P = 0$ ), as seen in the Fenland Study **(Extended Data Figure 3)**. We additionally identified a strong positive correlation of ISI with M/I ( $\rho = 0.522$ ,  $P = 5.10e-76$ ). Overall correlations were similar and consistent in direction between RISC and Fenland studies for primary overlapping traits of interest.

## 2. Genetic Discovery of post-challenge insulin resistance: European-only meta-analyses

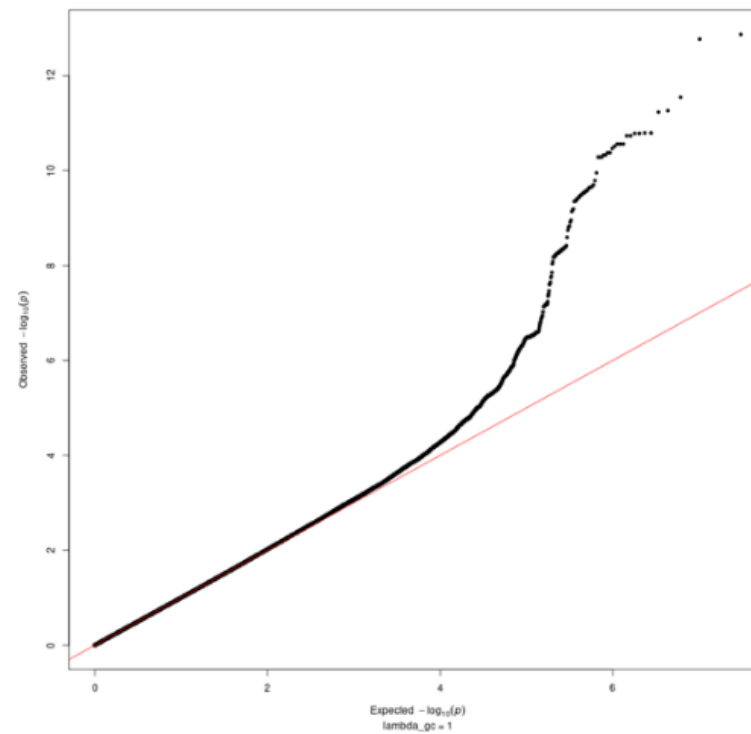
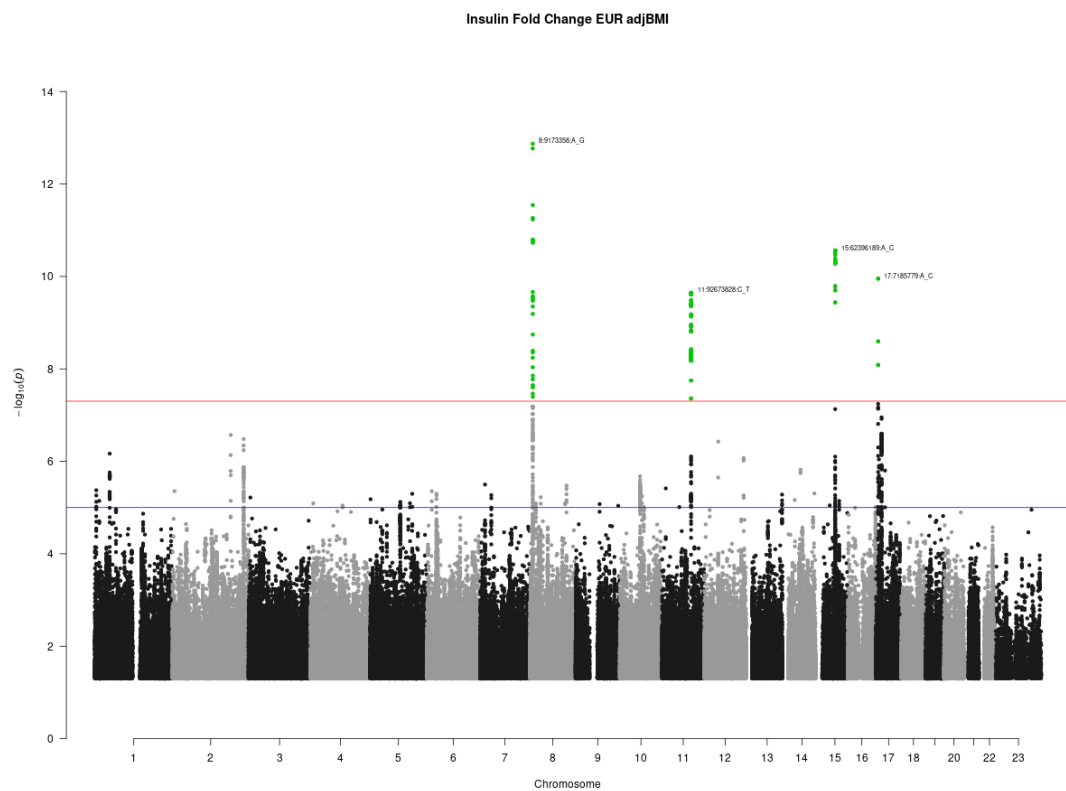
We performed fixed-effect meta-analyses within cohorts of European ancestry using METAL<sup>14</sup> for IFC and ISI. Manhattan and QQ plots summarising the results can be found in **Supplementary Figures 1-4** below. Unadjusted  $-\log_{10}$  P-values are indicated on the y axis of the Manhattan plots. QQ plots show the observed (Y) vs expected (X)  $-\log_{10}$  P value. Unadjusted P values are shown. In Manhattan plots on the left, the green points indicate those that meet genome wide significance. Red line indicates genome wide significance ( $P < 5e-8$ )\_ and blue line represents suggestive significance ( $P < 1e-5$ ).



**Supplementary Figure 1: Manhattan and QQ plots for European-only meta-analysis of Modified Stumvoll ISI adjusted for BMI**

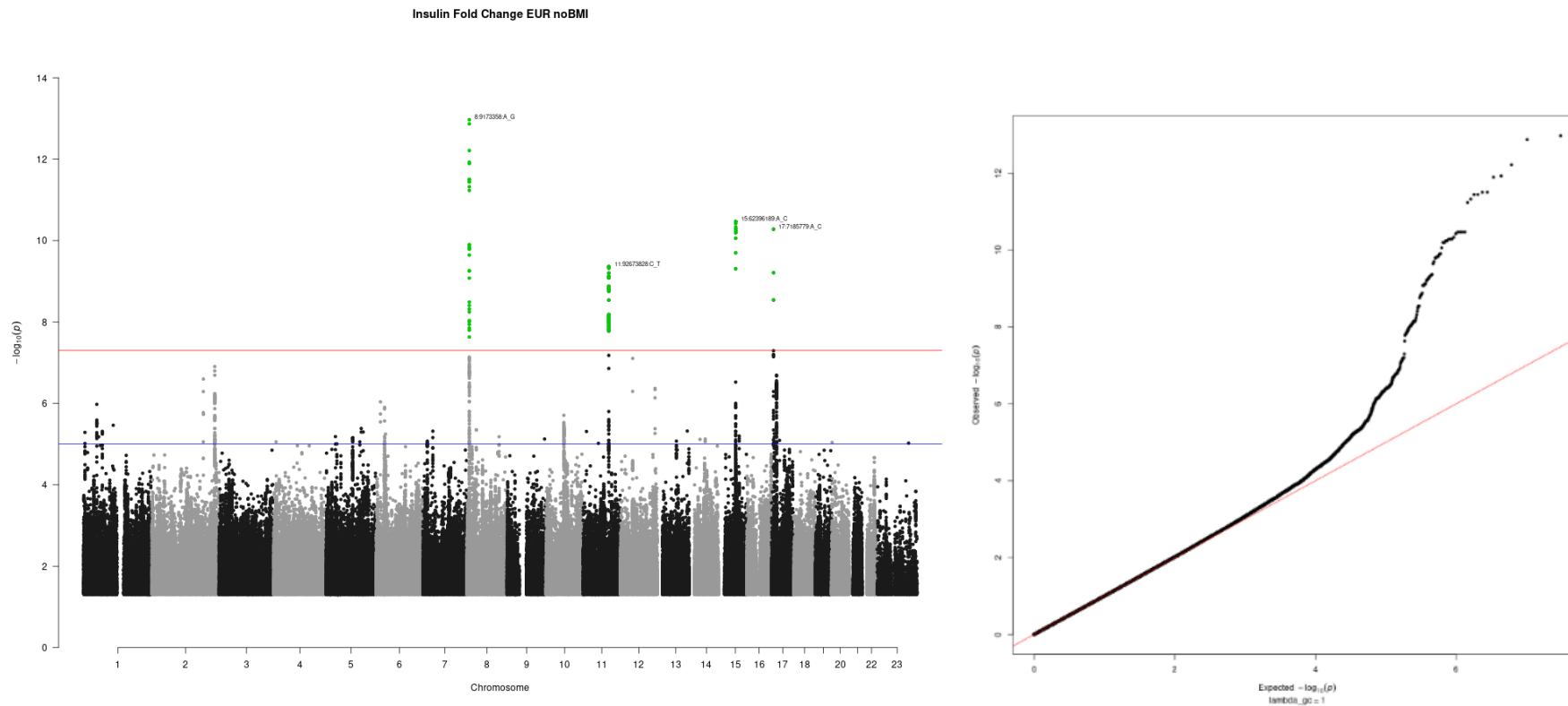


**Supplementary Figure 2: Manhattan and QQ plots for European-only meta-analysis of Modified Stumvoll ISI unadjusted for BMI**



**Supplementary Figure 3: Manhattan and QQ plots for European-only meta-analysis of insulin fold change adjusted for BMI**



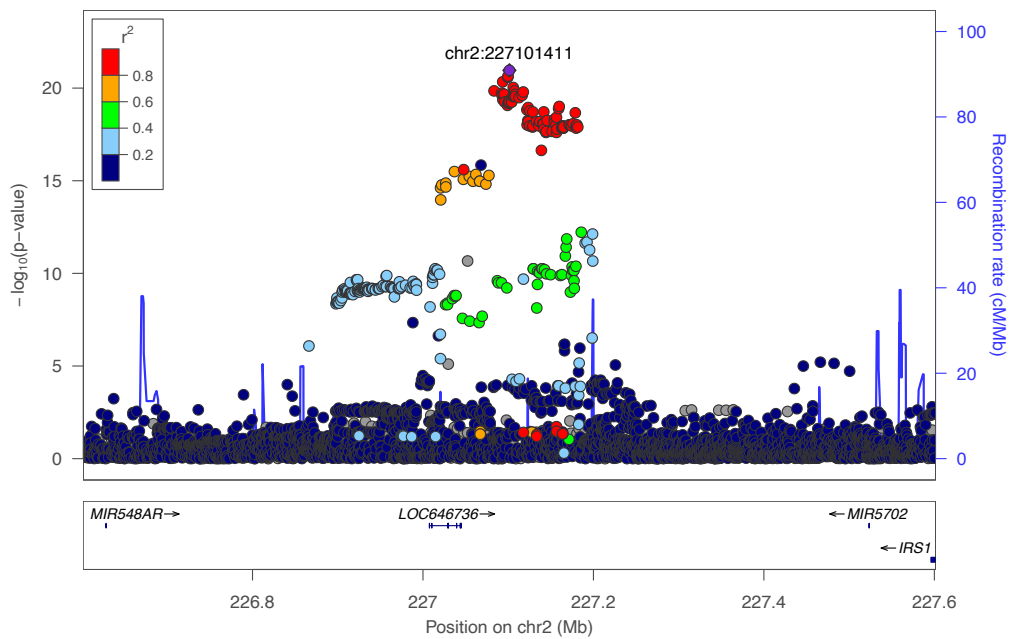


**Supplementary Figure 4: Manhattan and QQ plots for European-only meta-analysis of insulin fold change unadjusted for BMI**

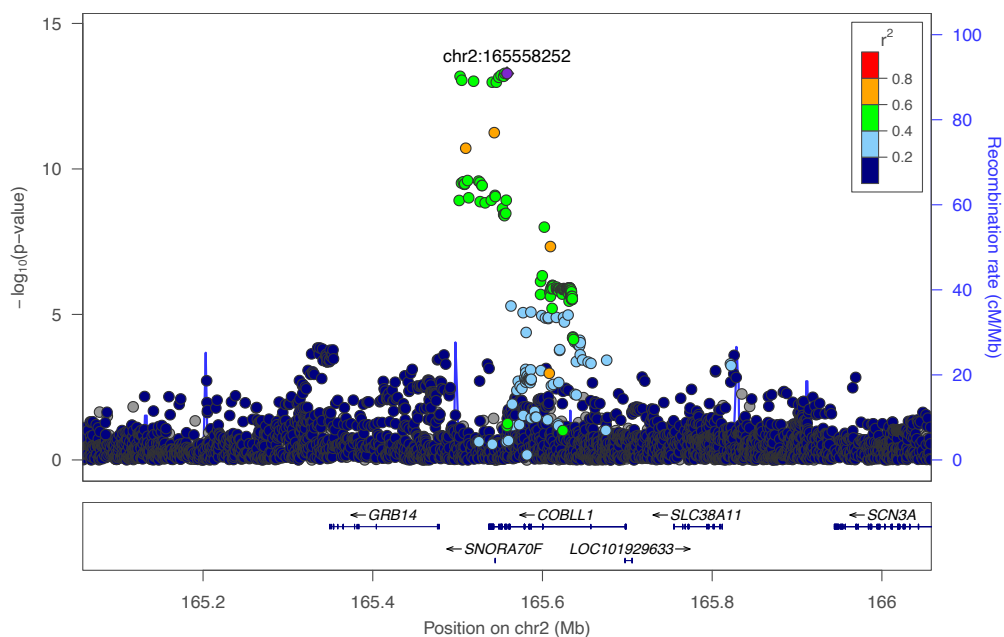
### 3. Regional plots of genetic loci associated with Modified Stumvoll ISI adjBMI (Supplementary Figure 5)

#### Supplementary Figure 5: Regional association plots of genetic loci associated with Modified Stumvoll ISI adjBMI in European-only meta-analysis.

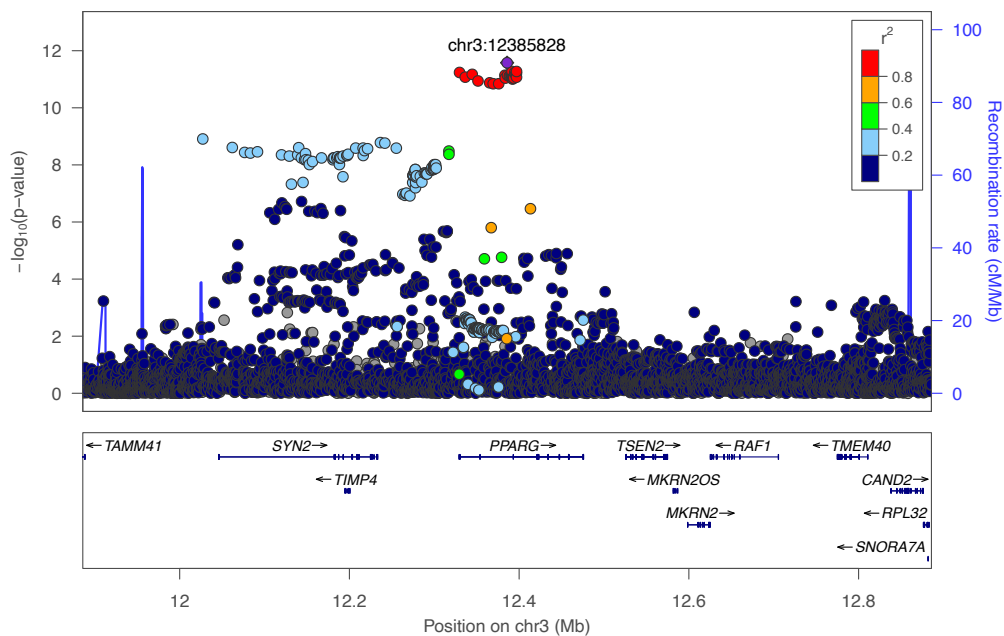
A locus was defined as  $\pm 500\text{kb}$  of the lead variant at a given locus. LD reference panel is 1000 Genomes CEU (European), the lead variant at a given locus is indicated by the purple triangle. Relative LD ( $R^2$ ) of other variants with this lead variants is indicated by the colour scale shown in the legend on the left hand side of the plot. Unadjusted  $-\log_{10}$  P-values are indicated on the y axis.



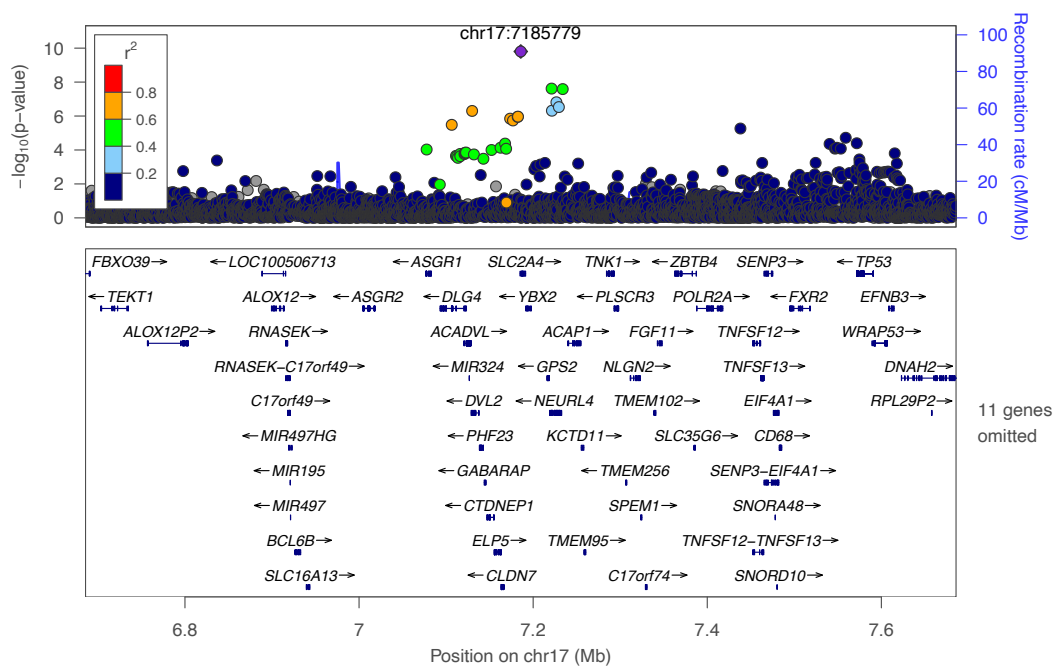
Supplementary Figure 5 a) IRS1



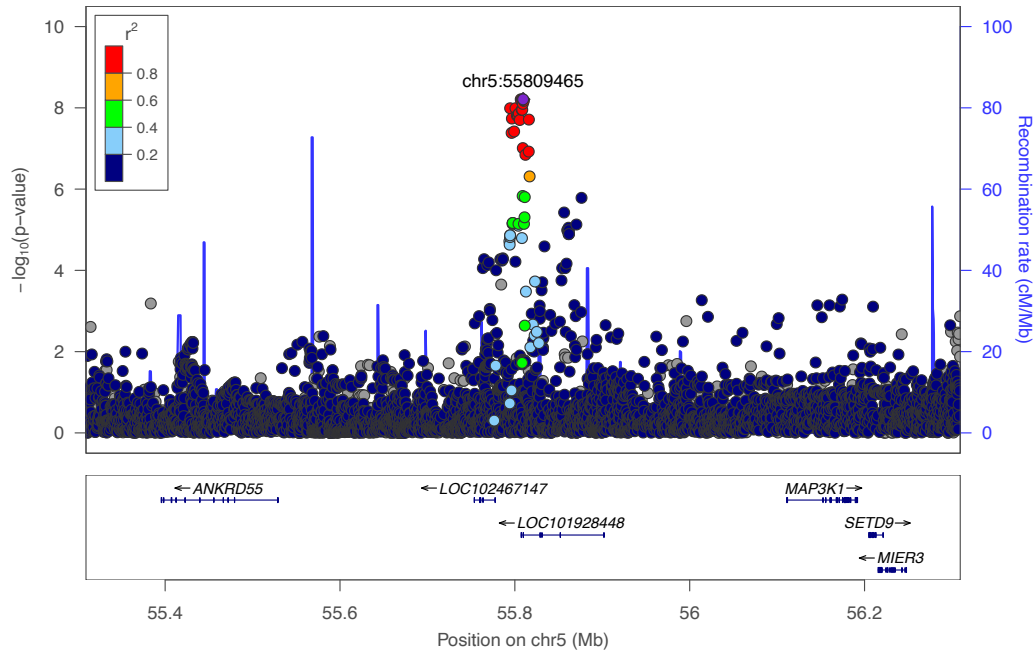
Supplementary Figure 5 b) COBLL1



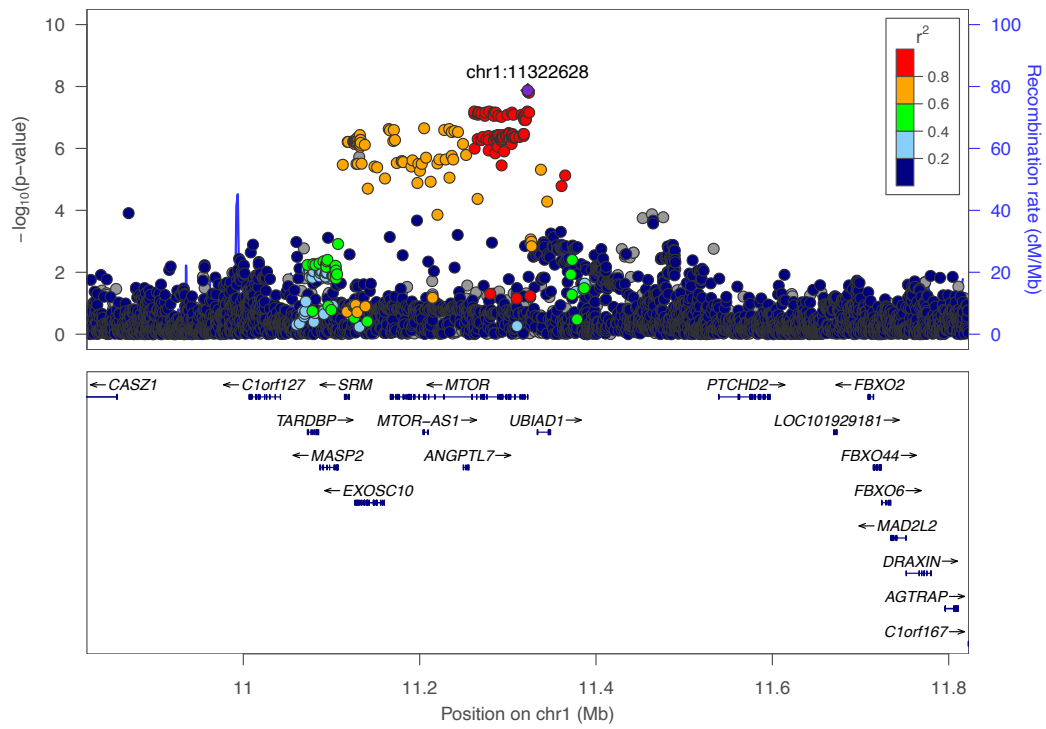
Supplementary Figure 5 c) PPARG



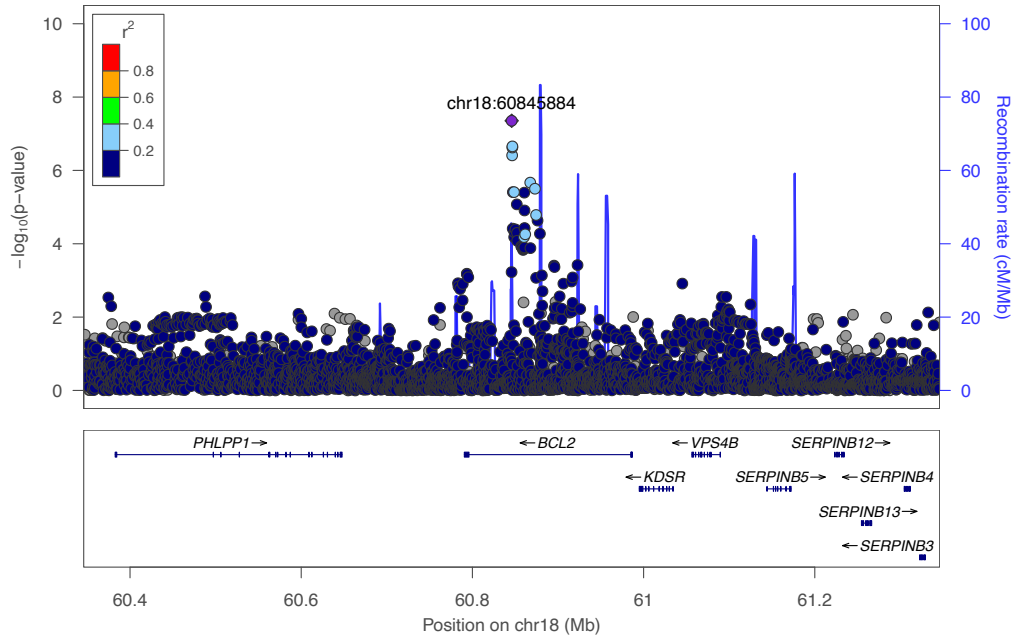
Supplementary Figure 5 d) SLC2A4



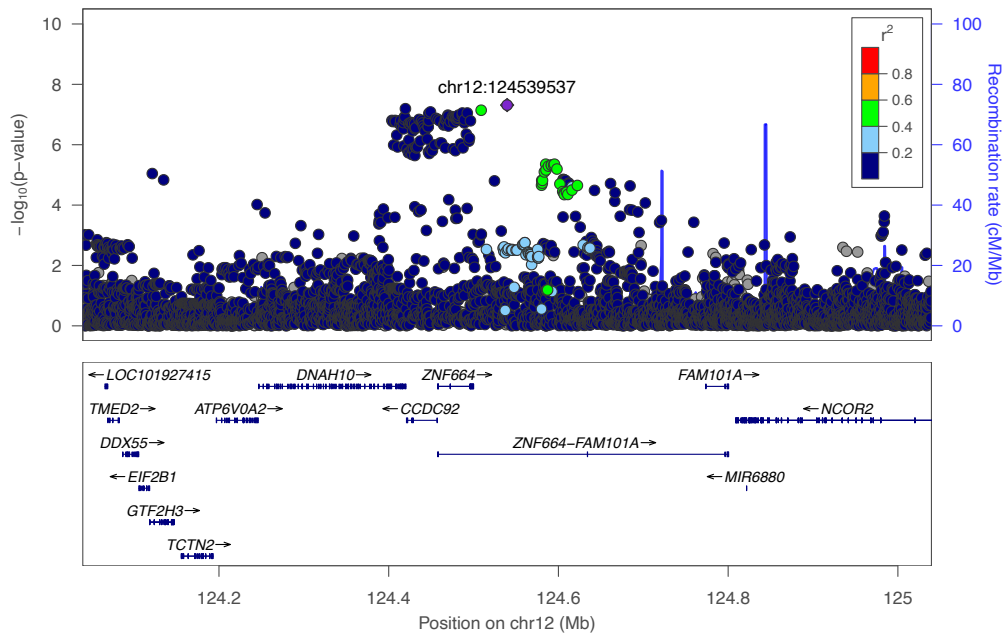
Supplementary Figure 5 e) C5orf67



Supplementary Figure 5 f) MTOR



Supplementary Figure 5 g) BCL2

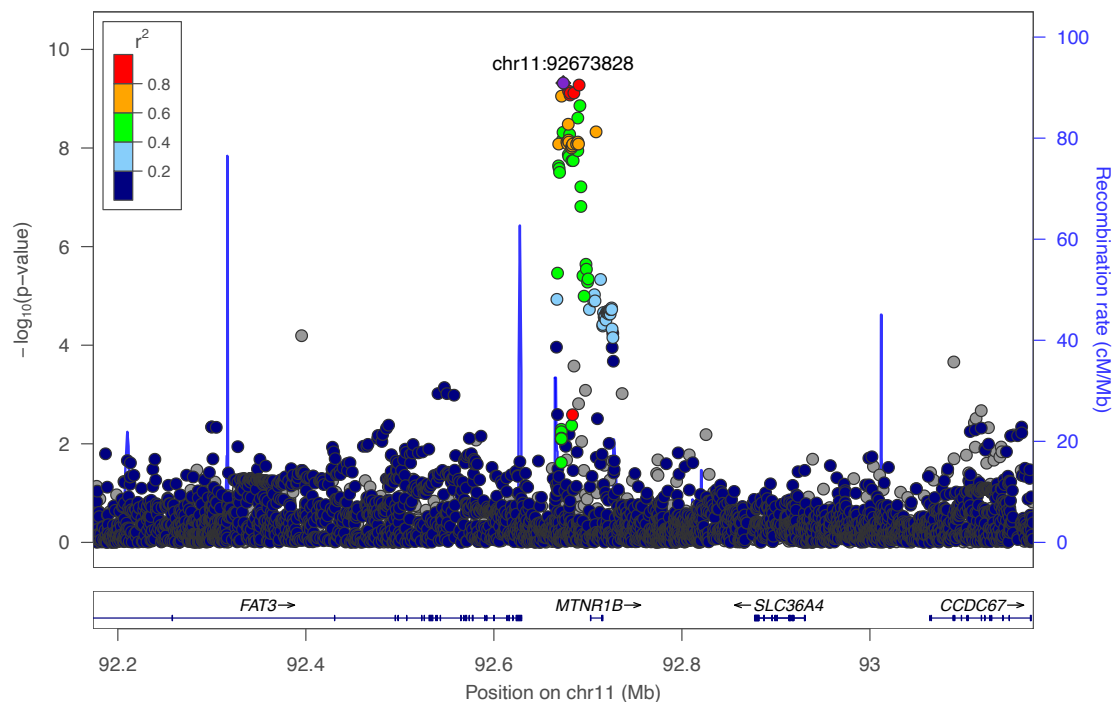


Supplementary Figure 5 h) FAM101A

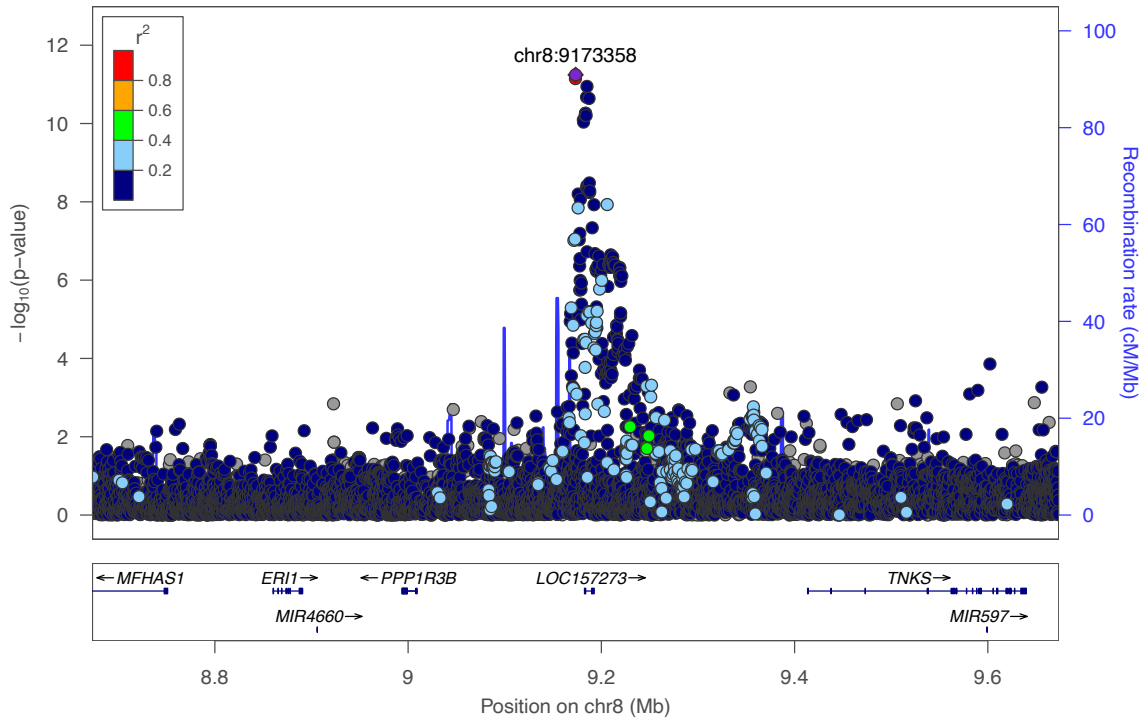
#### 4. Regional plots of genetic loci associated with Insulin Fold Change adjusted for BMI (Supplementary Figure 6)

Supplementary Figure 6: Regional association plots of genetic loci associated with insulin fold change adjBMI in European-only meta-analysis.

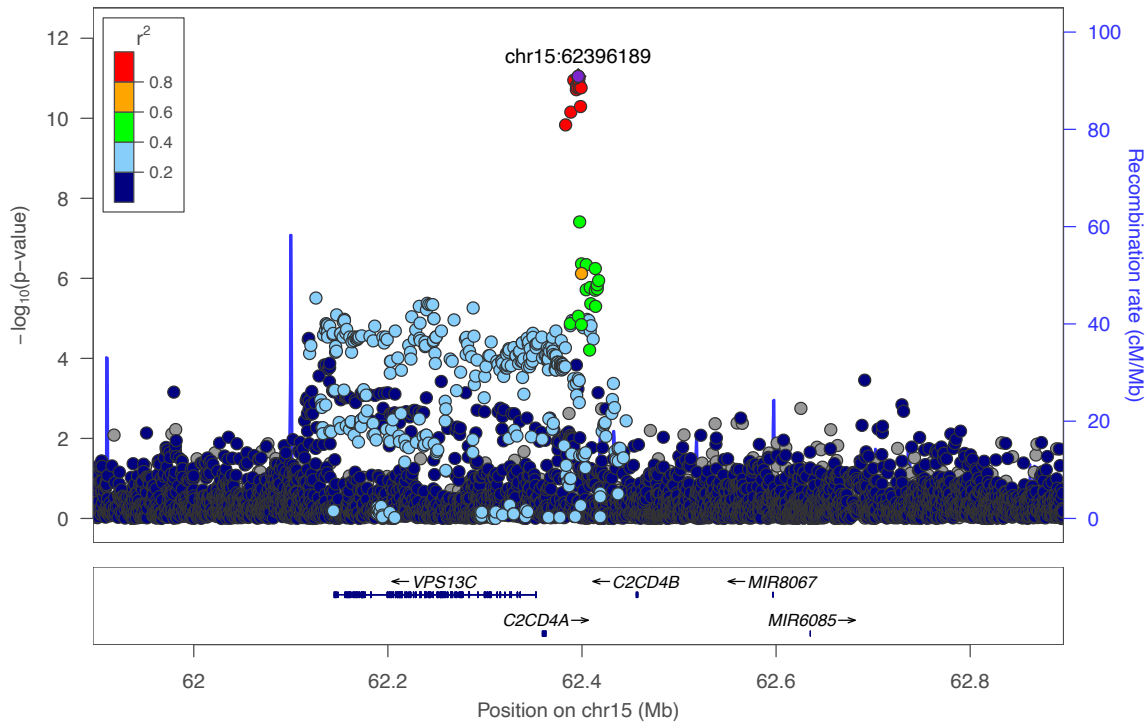
A locus was defined as  $\pm 500\text{kb}$  of the lead variant at a given, locus. LD reference panel is 1000 Genomes CEU (European), the lead variant at a given locus is indicated by the purple triangle. Relative LD ( $R^2$ ) of other variants with this lead variants is indicated by the colour scale shown in the legend on the left hand side of the plot. Unadjusted  $-\log_{10}$  P-values are indicated on the y axis. Unadjusted  $-\log_{10}$  P-values are indicated on the y axis.



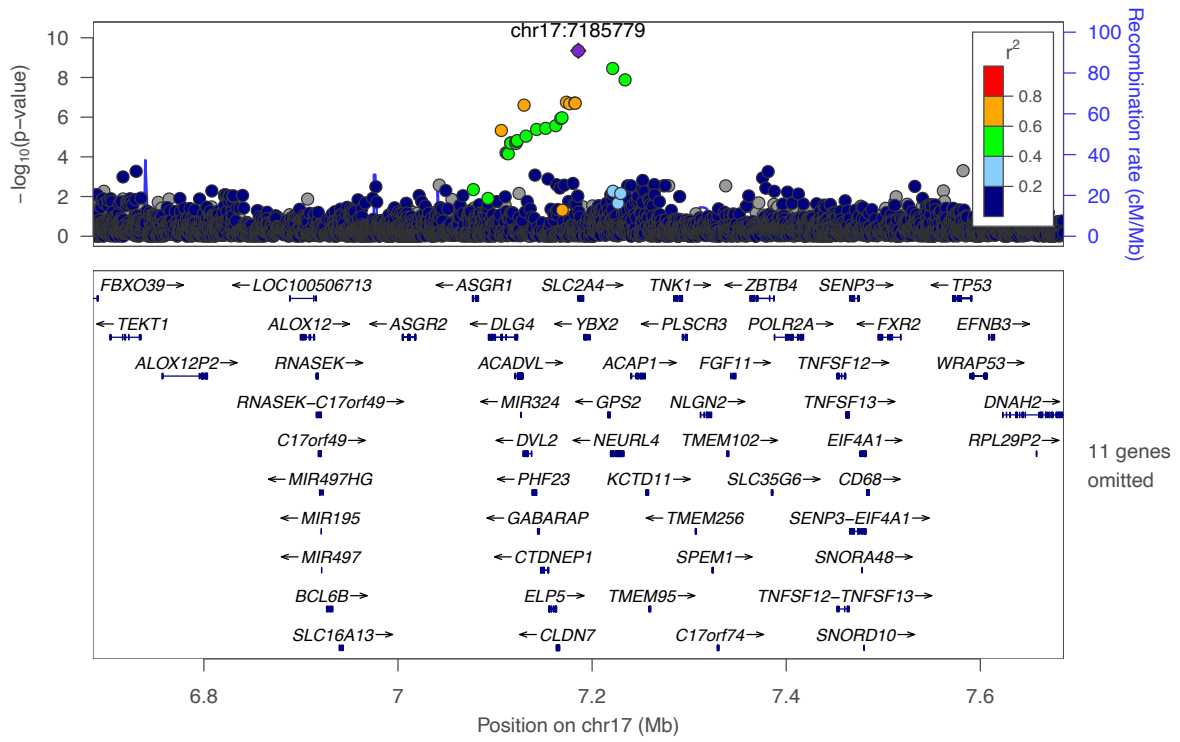
Supplementary Figure 6 a) MTNR1RB



Supplementary Figure 6 b) PPP1R3B



Supplementary Figure 6 c) C2CD4A

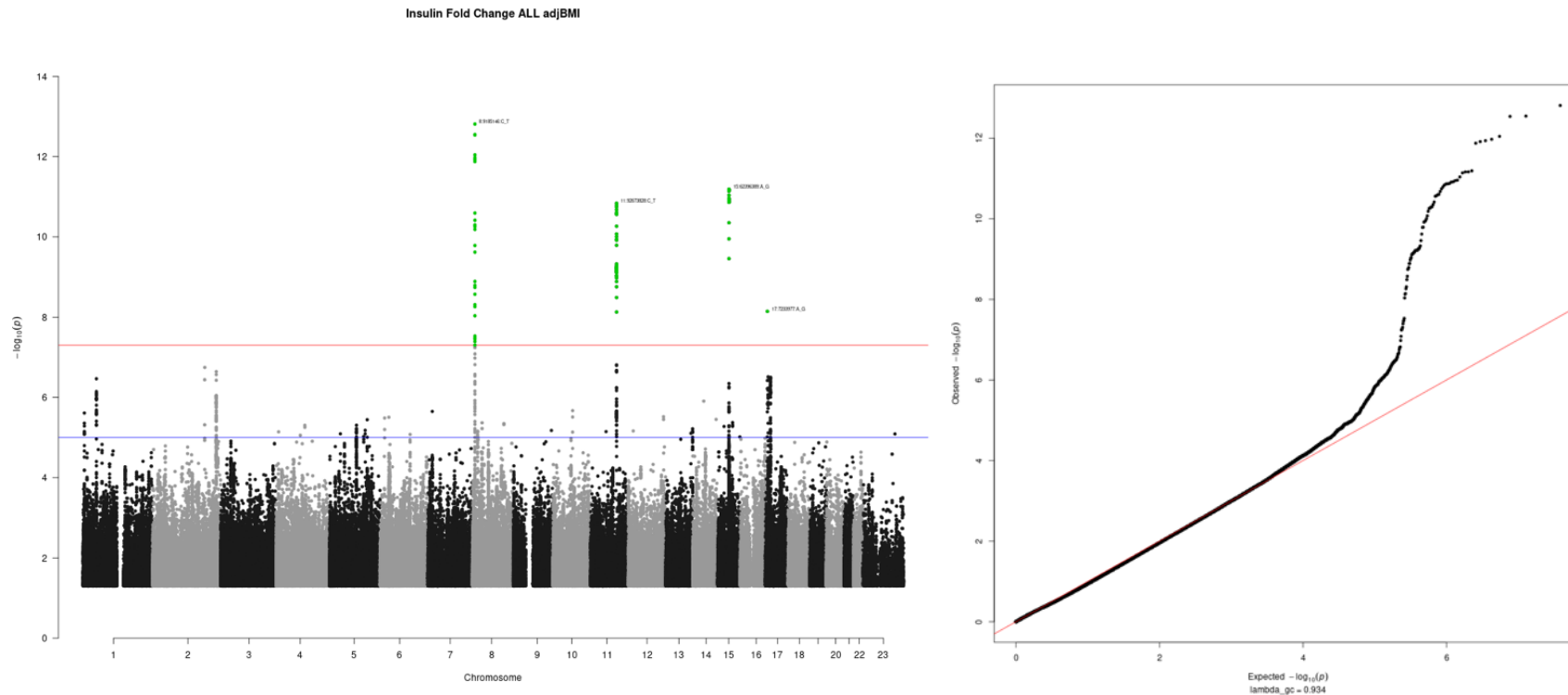


Supplementary Figure 6 d) SLC2A4

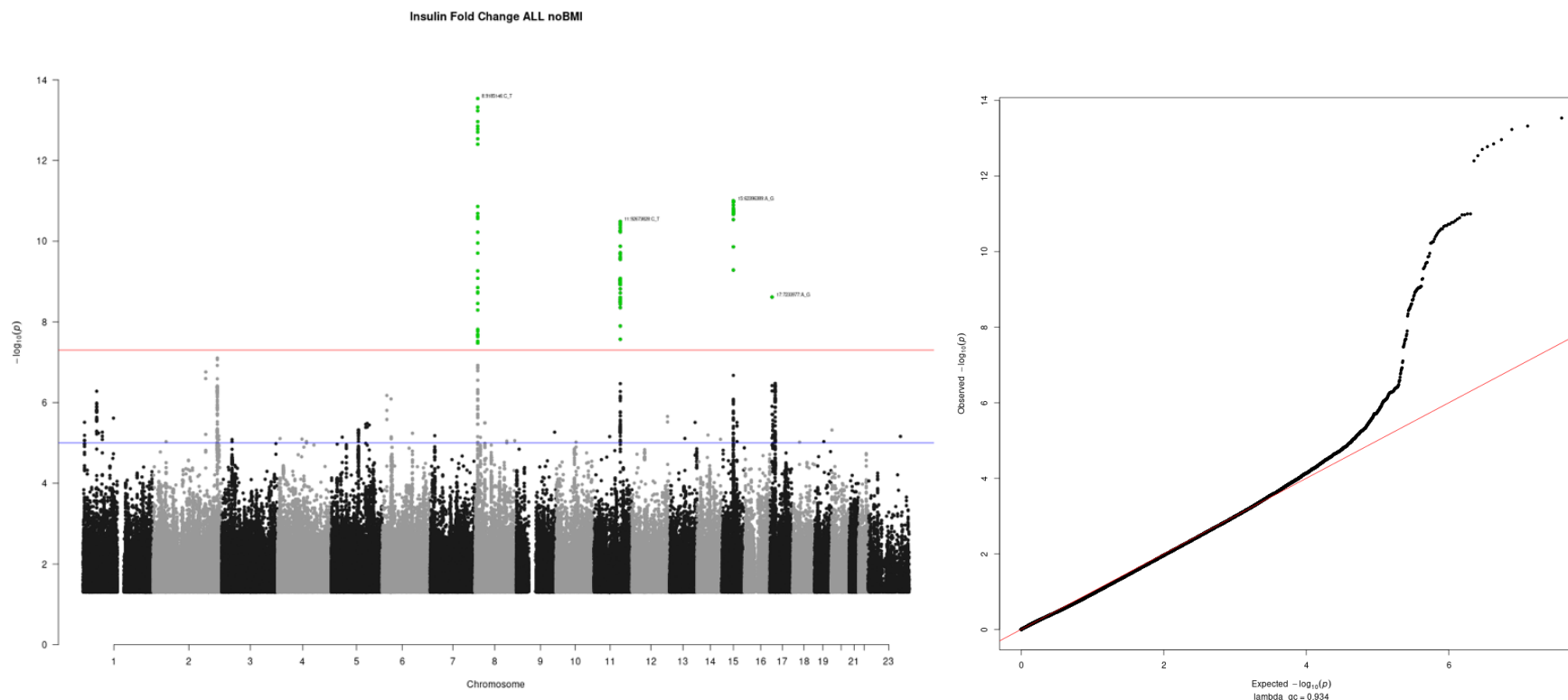


## 5. Genetic Discovery of post-challenge insulin resistance: non-European and multi-ancestry meta-analyses (Supplementary Figures 7-14)

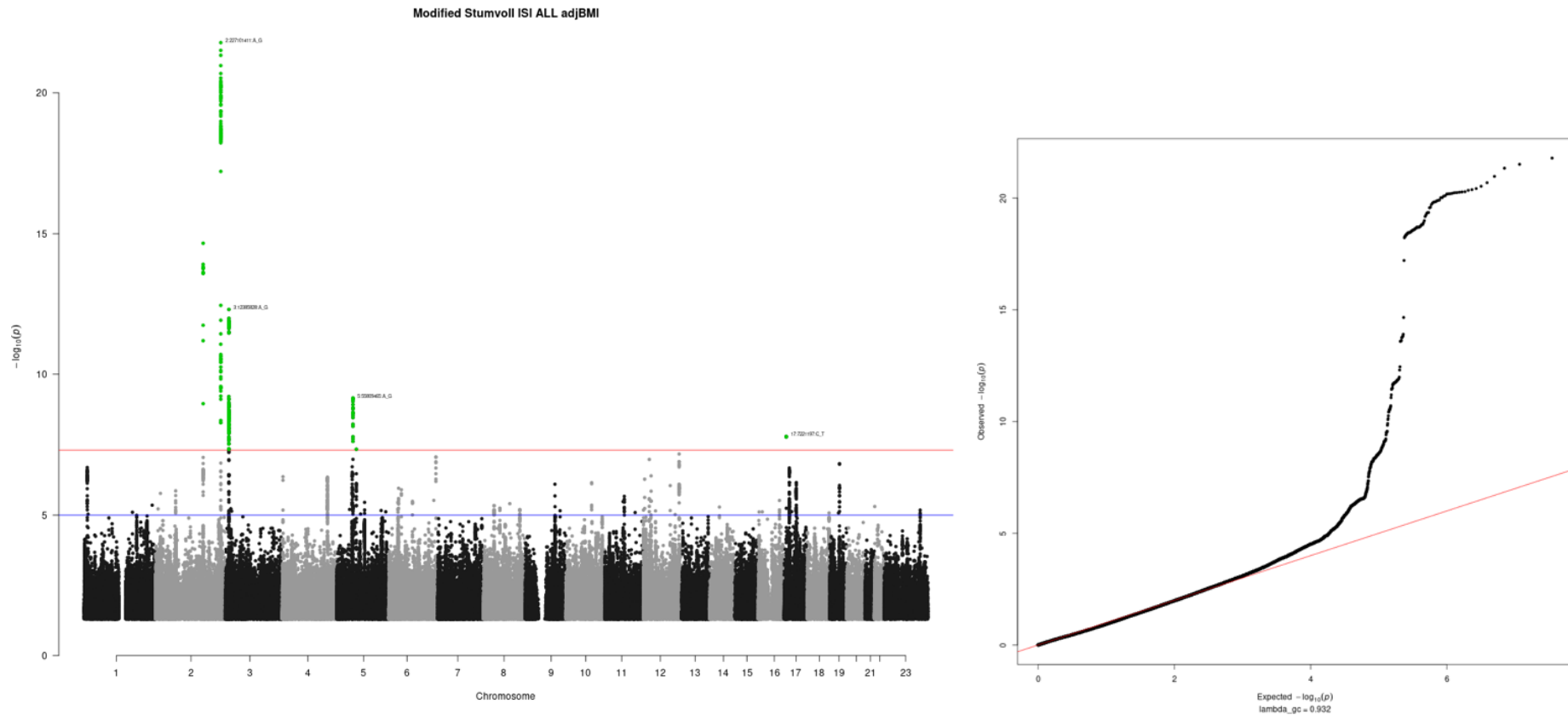
We performed fixed-effect meta-analyses separately within cohorts of Hispanic American ancestry using METAL<sup>14</sup> for IFC and ISI. To examine trans-ancestry effects we conducted random effects meta-analyses of these ancestry-specific results (EUR, EAS, HISAMR) using METAL (See **methods**). Manhattan and QQ plots summarising the results can be found in **Supplementary Figures 7-14** below. Unadjusted  $-\log_{10}$  P-values are indicated on the y axis of the Manhattan plots. QQ plots show the observed (Y) vs expected (X)  $-\log_{10}$  P value. Unadjusted P values are shown. In Manhattan plots on the left, the green points indicate those that meet genome wide significance. Red line indicates genomewide significance ( $P < 5e-8$ )\_ and blue line represents suggestive significance ( $P < 1e-5$ ).



Supplementary Figure 7: Manhattan and QQ plots for multi-ancestry (EUR, HISAMR, EAS) meta-analysis Insulin Fold Change adjusted for BMI

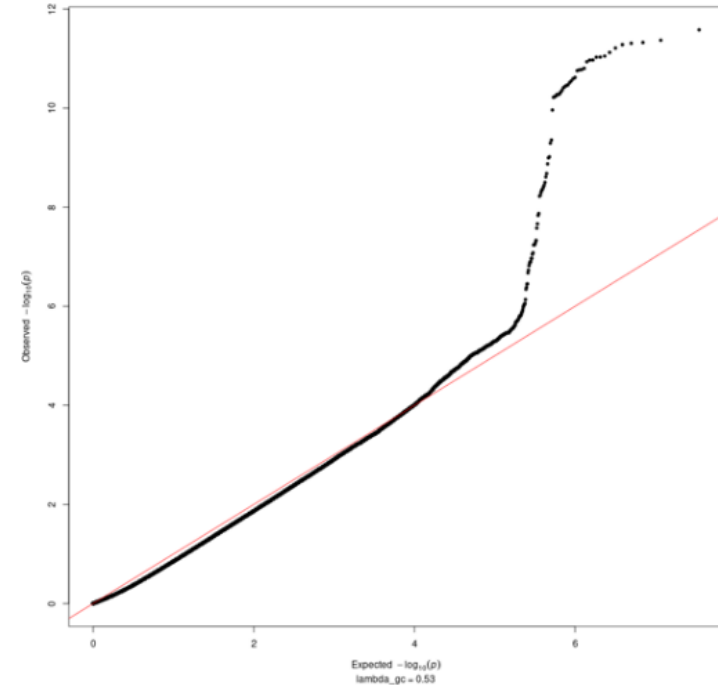
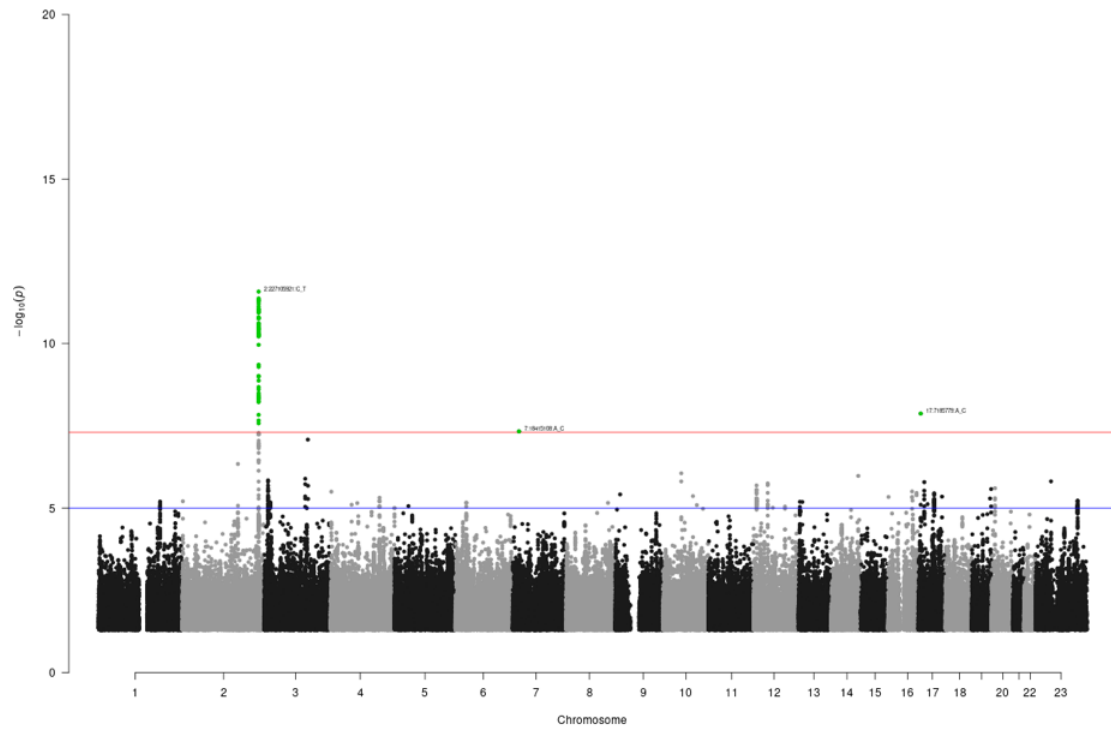


**Supplementary Figure 8: Manhattan and QQ plots for multi-ancestry (EUR, HISAMR, EAS) meta-analysis Insulin Fold Change unadjusted for BMI**



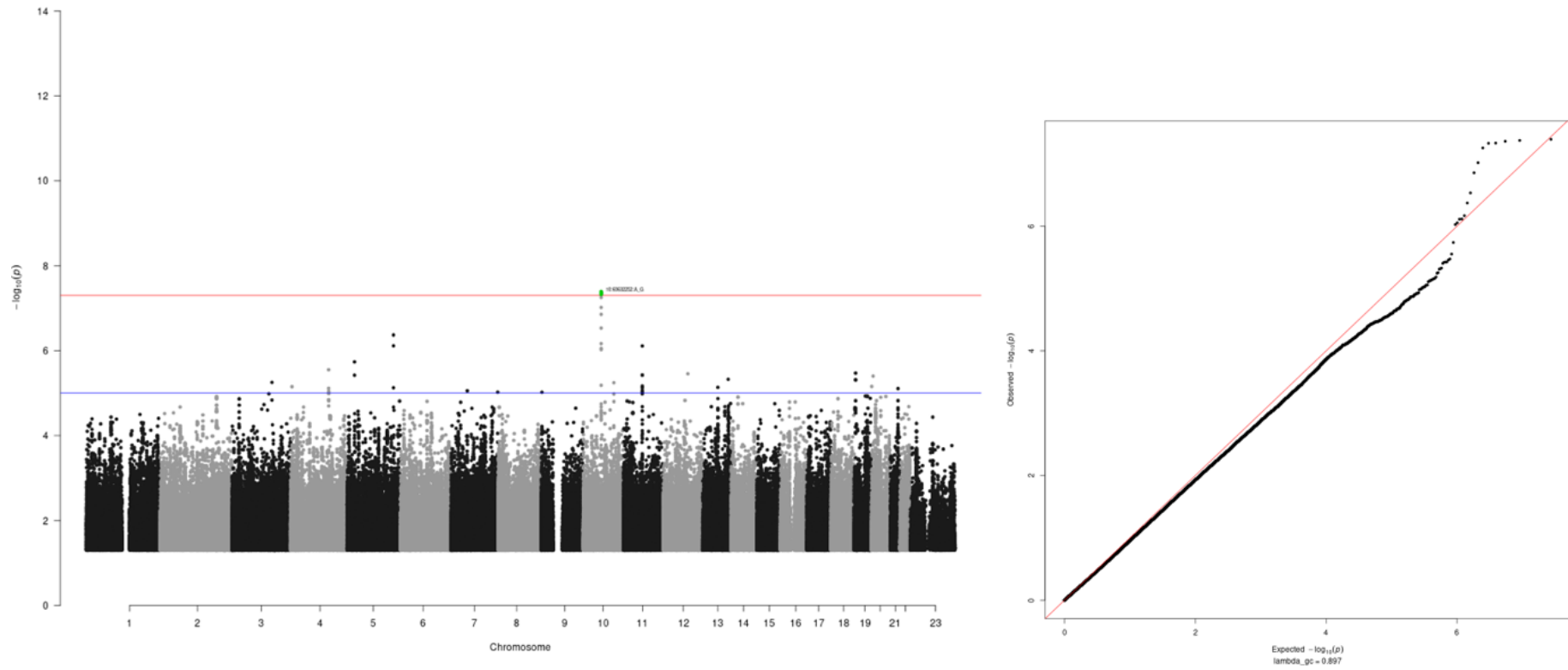
Supplementary Figure 9: Manhattan and QQ plots for multi-ancestry (EUR, HISAMR, EAS) meta-analysis Modified Stumvoll ISI adjusted for BMI

Modified Stumvoll ISI ALL noBMI



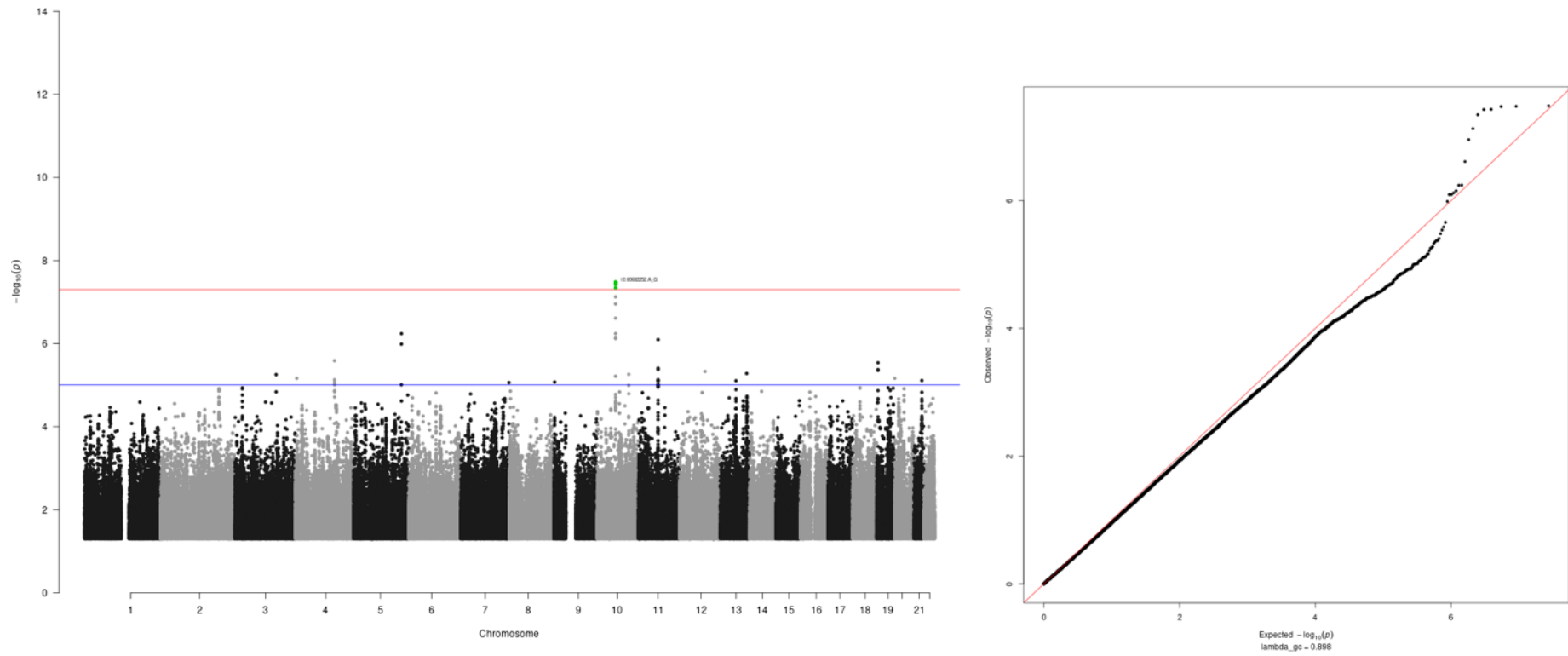
Supplementary Figure 10: Manhattan and QQ plots for multi-ancestry (EUR, HISAMR, EAS) meta-analysis Modified Stumvoll ISI adjusted for BMI

Insulin Fold Change NONEUR adjBMI



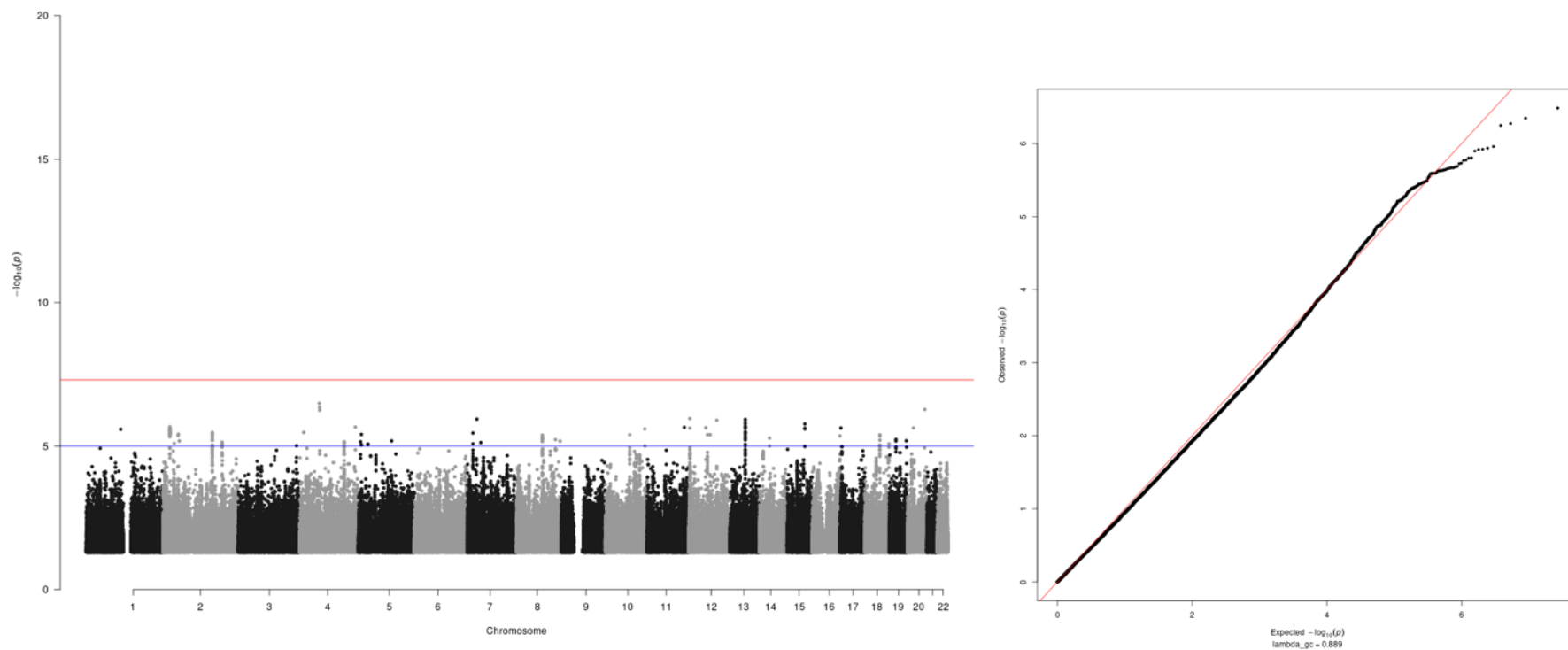
Supplementary Figure 11: Manhattan and QQ plots for non-European Ancestry (HISAMR, EAS) meta-analysis Insulin Fold Change adjusted for BMI

Insulin Fold Change NONEUR noBMI



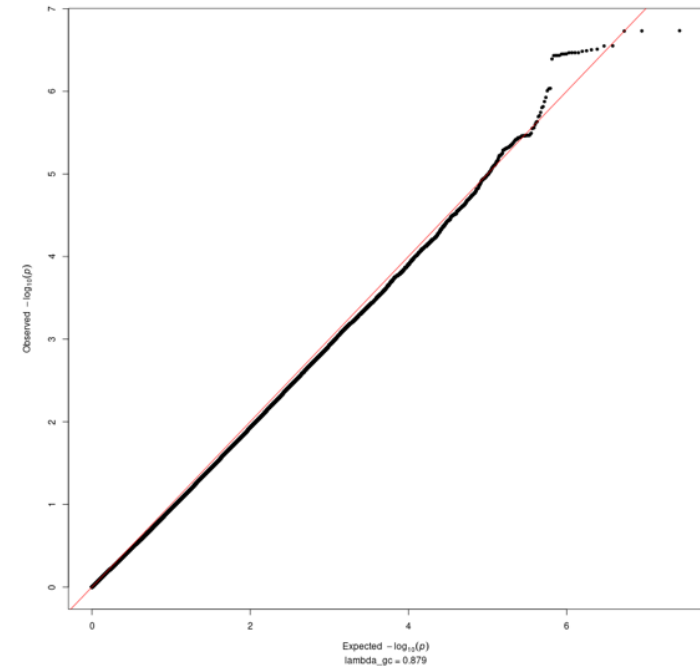
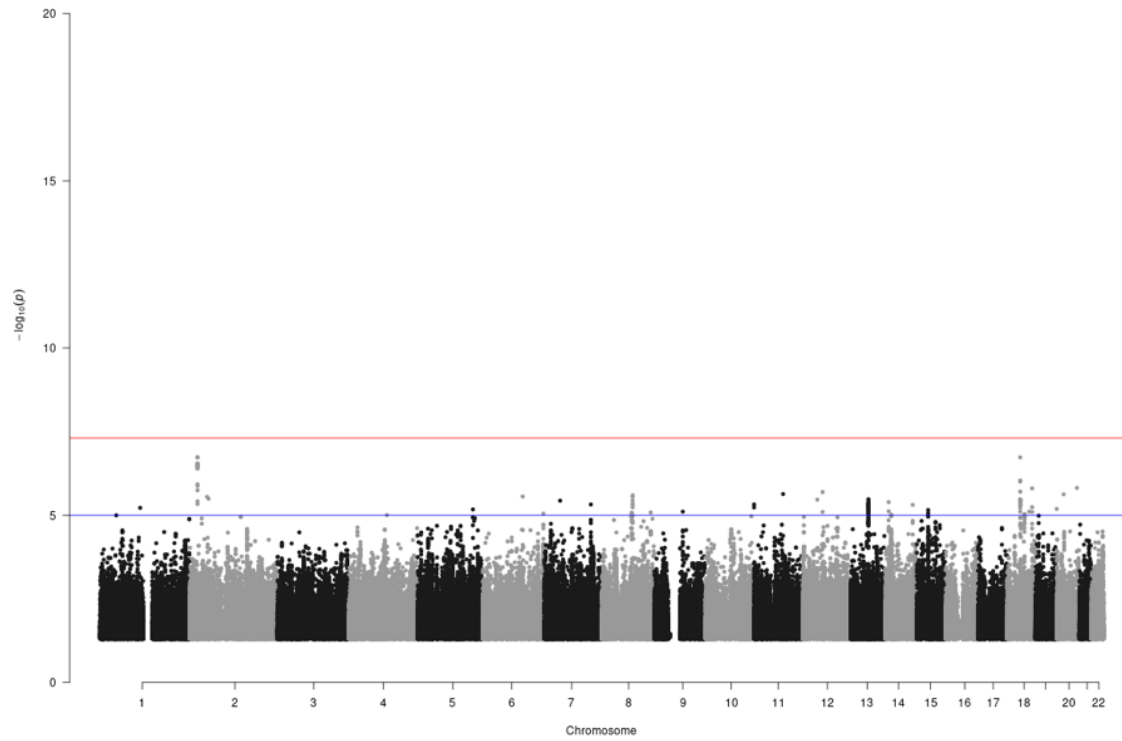
Supplementary Figure 12: Manhattan and QQ plots for non-European Ancestry (HISAMR, EAS) meta-analysis Insulin Fold Change unadjusted for BMI

Modified Stumvoll ISI NONEUR adjBMI



Supplementary Figure 13: Manhattan and QQ plots for non-European Ancestry (HISAMR, EAS) meta-analysis Modified Stumvoll ISI adjusted for BMI

Modified Stumvoll ISI NONEUR noBMI



Supplementary Figure 14: Manhattan and QQ plots for non-European Ancestry (HISAMR, EAS) meta-analysis Modified Stumvoll ISI unadjusted for BMI



## 6. rs60453193 (chr10:60632252\_A\_G (b37)) at *BICC1* is associated with Insulin Fold Change specifically in Non-European Cohorts

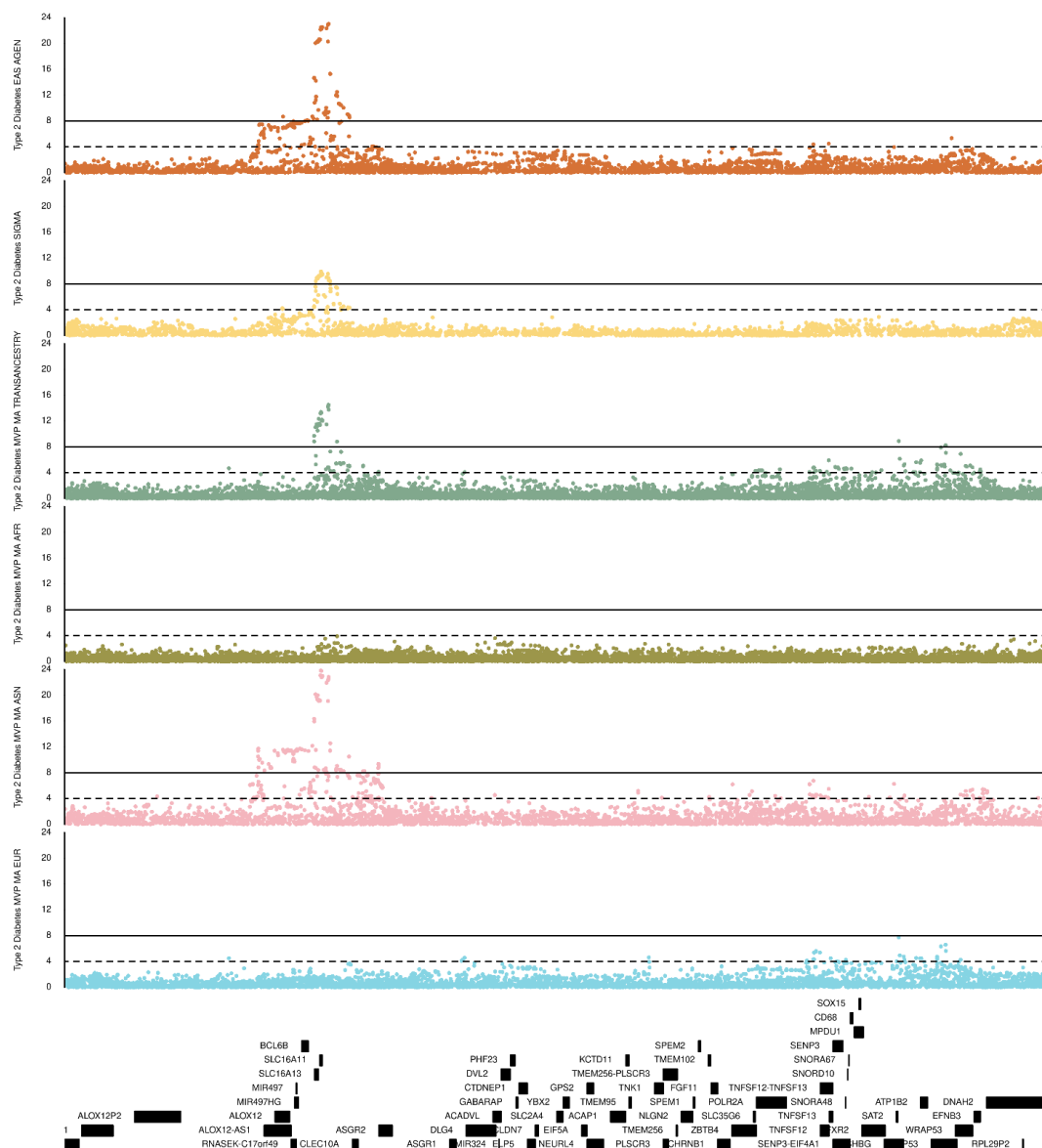
We additionally identified *BICC1* (rs60453193, beta = 0.43, SE = 0.08, P =  $4.06 \times 10^{-8}$ , N = 1,837) as a signal associated with IFC (**see Methods**). This locus has not been implicated in post-challenge insulin resistance previously. The effects were consistent in meta-analyses of only East Asian (beta = 0.406, SE = 0.204, P = 0.046, N = 739) and Hispanic American ancestry studies (beta = 0.435, SE = 0.085, P =  $3.15 \times 10^{-7}$ , N = 1,098) (**Supplementary Tables 6 and 7; Extended Data 5-6**). *BICC1* was not associated with IFC in meta-analysis of studies only of European ancestry (beta = -0.0026, SE = 0.010, P = 0.791, N = 50,671; **Supplementary Table 6**); despite being common in all ancestry groups (MAF European = 12%, Hispanic American = 7.8%, East Asian = 4.2%). Further, *BICC1* was not statistically significant in the multi-ancestry analyses of IFC, including all cohorts (beta = 0.26, SE = 0.18, P = 0.16, N = 52,508; **Supplementary Tables 6 and 7**), likely due to cohorts of European ancestry dominating the overall sample size (97%). Variants within this locus ( $\pm 500$ kb of rs60453193) have been reported to be associated with T2D at suggestive significance in multi-ancestry analyses<sup>8</sup> (Minimum P =  $1.0 \times 10^{-4.7}$ ), however rs60453193 itself is not associated (MVP multi-ancestry T2D: beta = 0.0079, SE = 0.0062, P = 0.020)<sup>8</sup>. Rare, damaging variants in *BICC1* have been reported to be nominally associated with T2D in individuals of European ancestry.<sup>15</sup>

## 7. SLC2A4 is in perfect D' with variants reported to be associated with type 2 diabetes in Hispanic American and East Asian ancestries.

### SLC2A4 locus and T2D risk

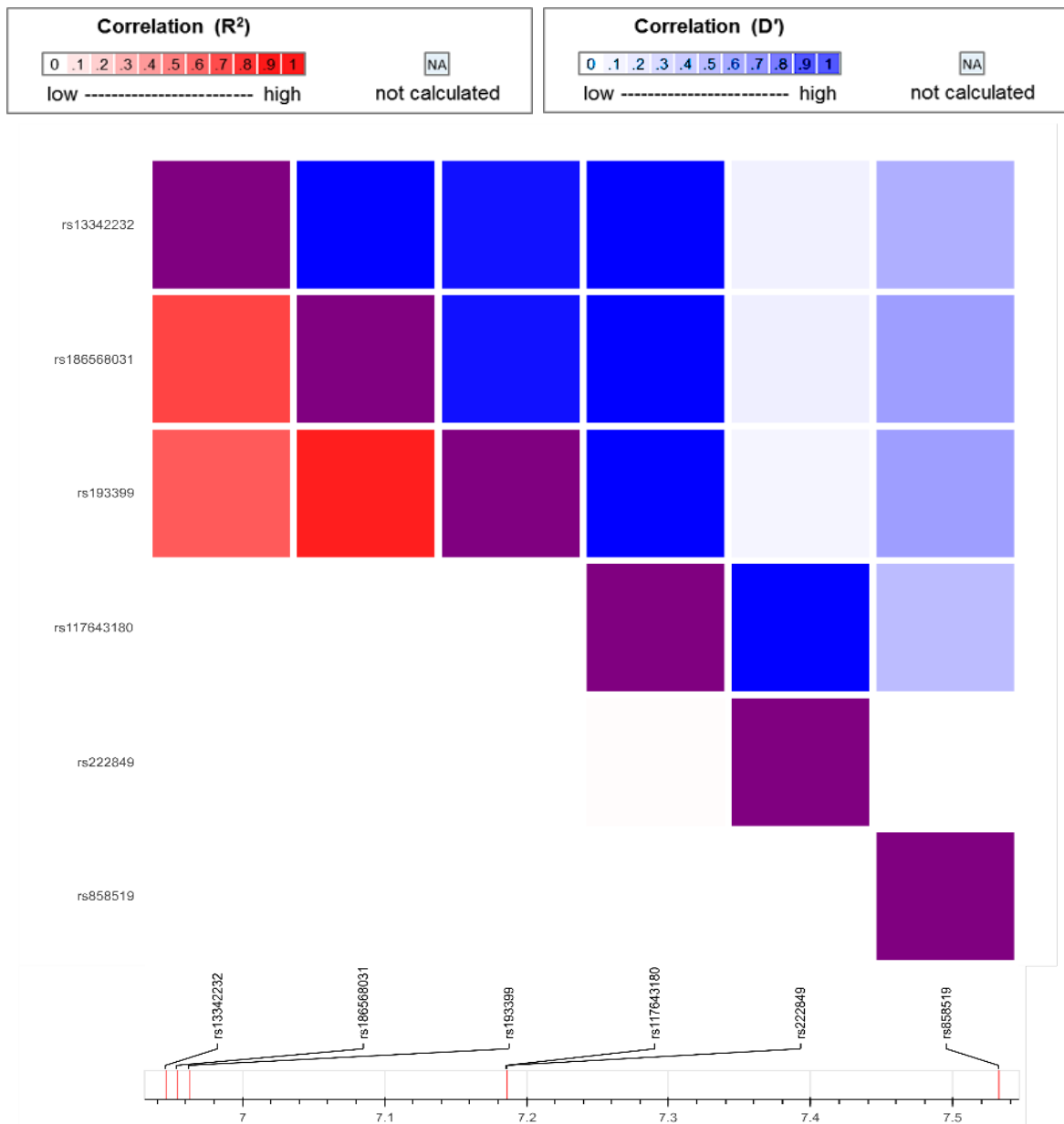
The lack of evidence of a shared signal between post-challenge insulin resistance and T2D at SLC2A4 in cohorts of European ancestry is surprising, since lead variant rs117643180 is located in the first intron of *SLC2A4*, encoding GLUT4, a key player in post-prandial glucose uptake. rs117643180 is also associated with higher 2 h glucose (**Supplementary Table 6**), one of the diagnostic criteria for T2D<sup>7,16</sup>. There are several reasons which may explain this. First the true causal variant for T2D may not be captured. *SLC2A4* lead variant rs117643180 is in perfect linkage disequilibrium (LD;  $D' = 1$ ) with lead SNPs at T2D loci identified in East Asian and Hispanic American ancestries (**Supplementary Figures 15- 17, Supplementary Table 18**).<sup>17,18</sup> The respective lead variants at this locus in each ancestry are too rare in other ancestries, such as European descent, to be included in previous efforts (**Supplementary Table 18**)<sup>19</sup>. The small sample size of non-European ancestries for post-challenge insulin resistance traits, unfortunately does not allow formal statistical colocalisation testing. Finally, it is possible that epistatic effects may play a role and the A risk allele of rs117643180 for post-challenge insulin resistance co-segregates with the established nearby protective T2D C allele of rs858519 (T2D association at rs858519: Beta = -0.025, SE = 0.004, P = 1.28e-9, EA = C EAF = 0.505)<sup>8</sup>. For instance in 1000 genomes<sup>20</sup>, the rs117643180-A risk allele is present at a frequency of 3.1% in European ancestry, and when present, this co-segregates on the same haplotype with the T2D protective C allele at rs858519 approximately 42% of the time<sup>21</sup>.

Finally our analyses focus on variants with MAF > 0.5% and there is suggestive evidence of rare predicted damaging missense variants (**see methods**) in *SLC2A4* being associated with T2D (MAF < 0.005%, N SNPs = 68) in ~450,000 European ancestry participants from the UK Biobank with whole exome sequencing<sup>15</sup>. These variants have been reported to be nominally associated with increased diabetes risk (UK Biobank field 20002 #1220, OR = 1.79 [1.25-2.68], P = 5.5x10<sup>-3</sup>), increased random glucose (Beta = 0.11, SE = 0.05, P = 0.03) and increased HbA1c (Beta = 0.13, SE = 0.06, P = 0.018).<sup>15</sup>



**Supplementary Figure 15: SLC2A4 locus regional association plot for Type 2 Diabetes across studies in different ancestry groups.**

Locus is defined as  $\pm 500\text{kb}$  from lead ISI and IFC SNP rs117643180. From top to bottom the tracks denote the following publicly available Type 2 Diabetes meta-analysis results: Track 1: East Asian Ancestry (AGEN; Spracklen et al; 2021)<sup>17</sup>. Track 2: Hispanic American ancestry, SIGMA consortium (Williams et al 2013)<sup>18</sup>. Track 3: MVP meta-analysis (Vujkovic et al; 2020)<sup>8</sup> trans-ancestry. Track 4: MVP meta-analysis (Vujkovic et al; 2020) African ancestry. Track 5: MVP meta-analysis (Vujkovic et al; 2020) Asian ancestry. Track 6: MVP meta-analysis (Vujkovic et al; 2020) European ancestry. Track 7: genes located at this locus. Y axis represents the  $-\log_{10}(\text{p-value})$  of association. (unadjusted p-value). The solid line refers to genome-wide significance threshold of  $P = 5e-8$ , and the dashed line denotes to suggestive significance threshold of  $P = 1e-5$ .



**Supplementary Figure 16: LD matrix for lead SNPs in IFC, ISI and T2D GWAS at SLC2A4 locus in European ancestry**

LD matrix generated for individuals of European ancestry using LDLink<sup>21</sup>. LD as denoted by  $R^2$  is shown in red, and  $D'$  in blue. The colour scales are indicated at the top of the plot.



## 8. Genetic associations of IFC and ISI with other cardiometabolic traits

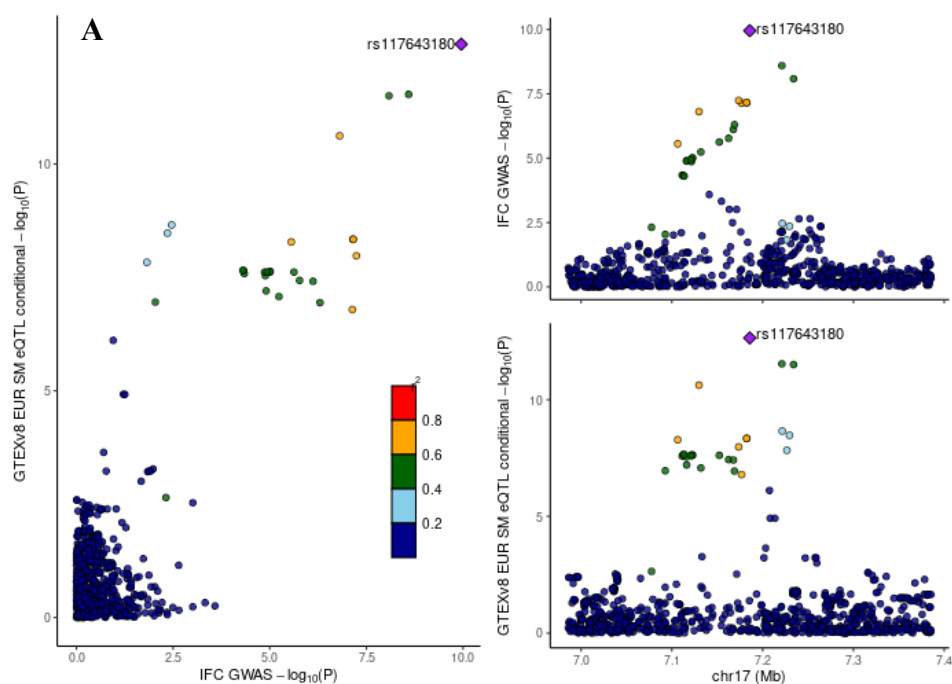
We further assessed the associations of increased genetic risk for post-challenge insulin resistance, represented by genetic risk scores (GRS; **see methods**) for ISI and IFC, with T2D related traits in unrelated European ancestry participants in UK Biobank (UKBB). We identified highly significant associations between both the ISI and IFC GRS with type 2 diabetes, with directions of associations consistent with post-challenge insulin resistance and increased risk of T2D (ISladjBMI: beta = -0.28, SE = 0.020, P =  $1.97 \times 10^{-43}$ ; IFCadjBMI: beta = 0.17, SE = 0.027, P =  $1.40 \times 10^{-10}$ ). Directionally consistent effects were also seen for GRS for ISI and IFC with HbA1C, random glucose, WHR adjusted for BMI, which were all increased with increased genetic risk of post-challenge insulin resistance (**Supplementary Table 13**). We further assessed the associations between GRS for ISI and IFC and glycaemic traits not available in UKBB in a subset of the Fenland Study. In this much smaller study (N max = 8,925), the IFC and ISI GRS were still nominally (P < 0.5) associated with dynamic glycaemic traits and with associations being directionally consistent with post-challenge insulin resistance (**Supplementary Table 14**).

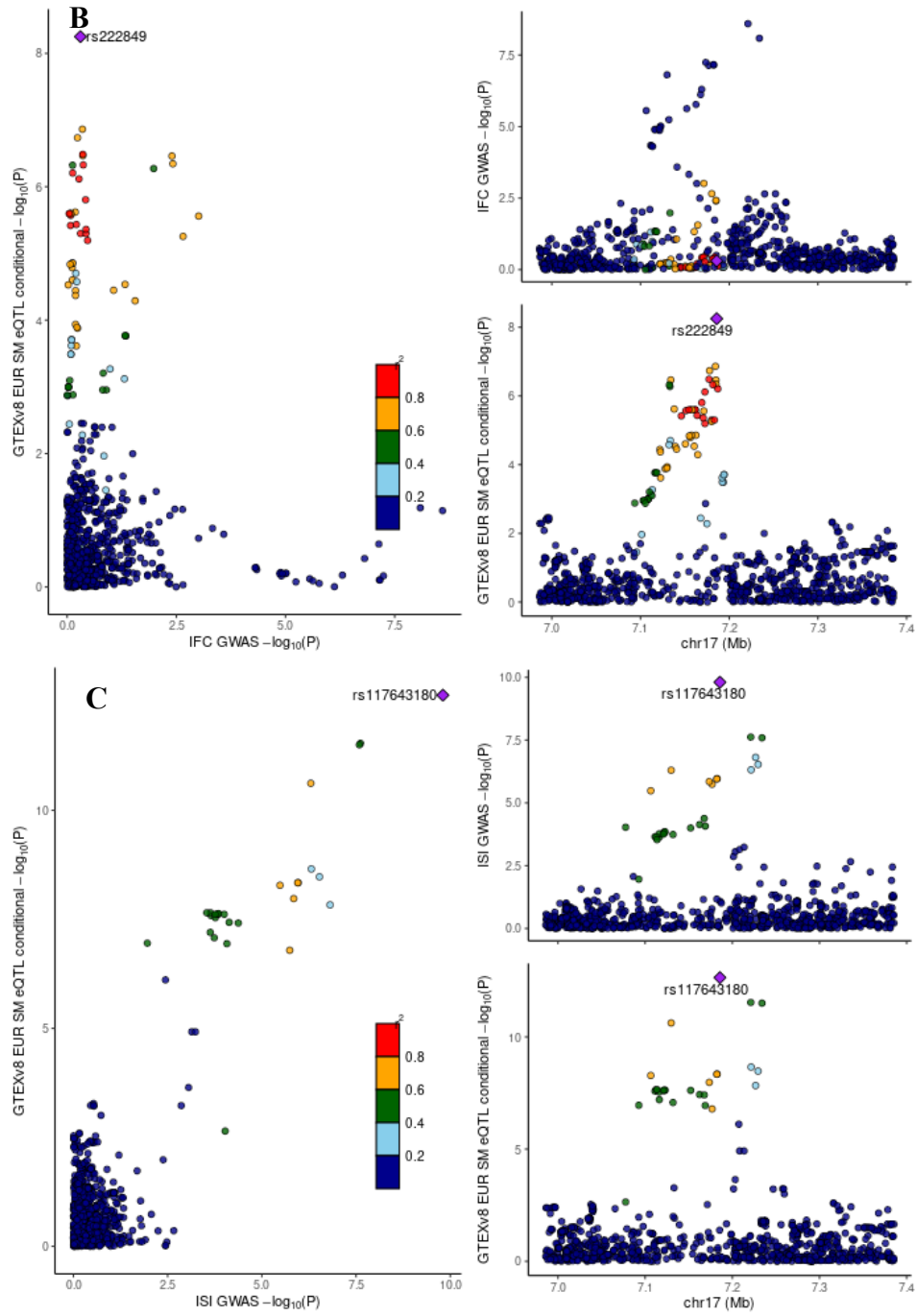
However, it should be noted that the Fenland Study is 1 of 28 studies that contributed to the GWAS meta-analyses from which these scores were constructed, reflecting 19% of the total sample size included in the discovery.

We used LD score regression analyses to assess genome-wide genetic correlations of ISI and IFC with a wider range of related biochemical and cardiometabolic traits. We identified strong genetic correlations of both ISI and IFC with 2h glucose, consistent with insulin resistance and the observational correlations in the Fenland Study (IFC and 2 h glucose: genetic correlation (rg) = 0.713, SE = 0.089, P =  $2.63 \times 10^{-22}$ ; ISI and 2 h glucose: rg = -0.820, SE = 0.048, P =  $2.12 \times 10^{-22}$ ; **Supplementary Tables 4, 15 and 16**). In contrast to ISI, which was strongly correlated with fasting insulin levels, IFC showed only a weak correlation, highlighting greater specificity as an indicator of post-challenge insulin resistance. Both ISI and IFC showed only moderate genetic correlations with T2D, consistent with the fact that multiple pathways contribute to its genetic architecture, with postprandial insulin resistance being only one of several aetiologies (**Supplementary Table 15 and 16**).

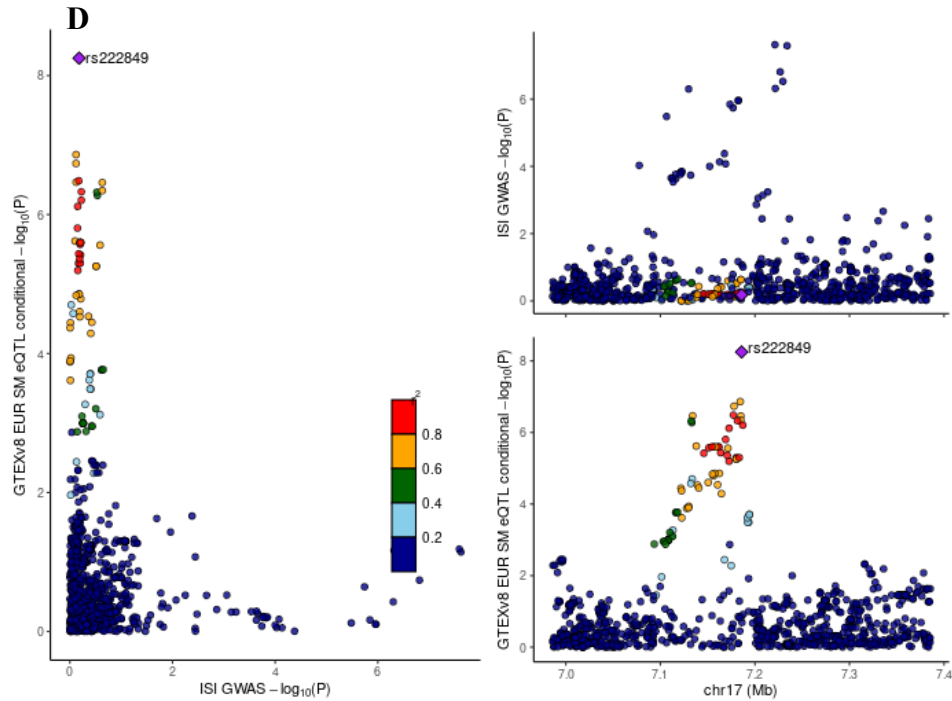
## 9. Colocalisation of post-challenge insulin resistance and eQTL for SLC2A4 in skeletal muscle at the SLC2A4 locus

The *SLC2A4* locus was the only locus where we did not see evidence of colocalisation of IFC/ISI with marginal statistics for the identified eQTL signal. This is due to two independent eQTL signals in skeletal muscle having been identified for *SLC2A4* at the *SLC2A4* locus (rs117643180)<sup>22</sup>. We ran conditional analyses using GCTA-cojo to generate conditional summary statistics conditioning on the primary (rs117643180) and secondary (rs222849) signals at this locus. Genetic of IFC/ISI and this eQTL signal were identified only when conditioning on the secondary signal (rs222849) showing that rs117643180 is the primary signal implicating both post-challenge insulin resistance and expression of *SLC2A4* at this locus (**Supplementary Table 17**).





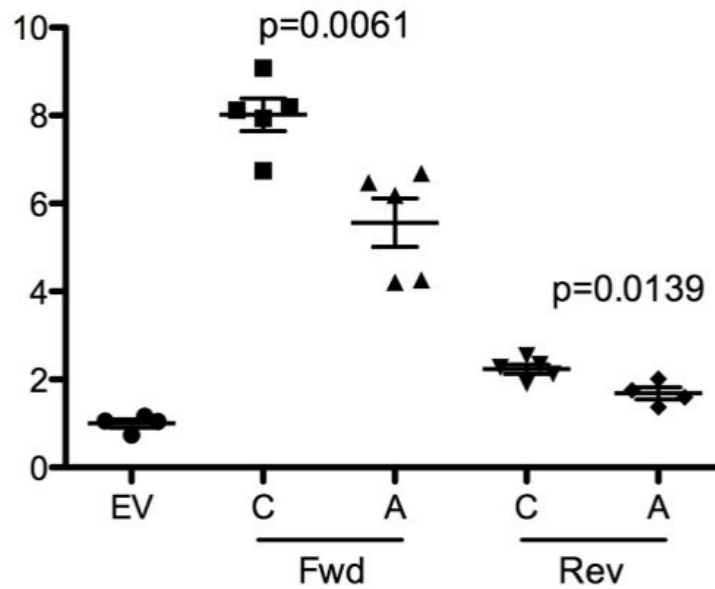




**Supplementary Figure 17: statistical colocalisation of ISI/IFC with conditional summary statistics for eQTL in skeletal muscle at the SLC2A4 locus.**

A) IFC and SLC2A4 eQTL statistics conditioned on the secondary signal rs222849. B) IFC and SLC2A4 eQTL statistics conditioned on the primary signal rs117643180. C) ISI and SLC2A4 eQTL statistics conditioned on the secondary signal rs222849. D) ISI and SLC2A4 eQTL statistics conditioned on the primary signal rs117643180. Locus was defined as  $\pm 200\text{kb}$  of rs117643180 for colocalisation analyses. The colour of points, as indicated in the legend, represent  $R^2$  values compared to the lead SNP (1000 genomes European LD reference) for each signal annotated with a purple diamond. Unadjusted  $-\log_{10}$  p-values are displayed.

**10. rs117643180 (*SLC2A4*) affects expression of GLUT4 in skeletal muscle through changes in transcriptional regulation**



**Supplementary Figure 18: rs117643180 exhibits allelic differences in transcriptional activity.**

229-bp fragments flanking rs117643180-C or rs117643180-A were cloned upstream of a minimal promoter driving luciferase expression in the forward and reverse orientations with respect to the promoter. Values represent fold-change of firefly luciferase/*Renilla* activity normalized to empty pGL4.23 vector in undifferentiated LHCN-M2 myoblasts. Error bars represent the SEM of four or five independent clones tested in duplicate wells, individual points represent independent clones. P-values are calculated from two-sided t-tests.

## **11. Tissue of Action at post-challenge insulin reflects loci reflect tissues implicated in post-prandial insulin action.**

To delineate potential tissue of action for post-challenge loci we employed LDSC-SEG to identify if the heritability of IFC or Modified Stumvoll ISI is enriched in regions surrounding genes with the highest specific tissue or cell type expression.

Modified Stumvoll ISI showed nominal genome-wide enrichment of active chromatin state in adipose, liver, skeletal muscle and pancreatic tissues, as well as nominal enrichment in tissue-specific gene expression in liver, muscle tissues and other T2D unspecific tissues (**Supplementary Tables 24 and 25**). Insulin fold change showed nominal enrichment in active chromatin marks in the liver, pancreas, and other tissues, as well as nominal enrichment in liver-specific gene expression. However, none of these annotations survives multiple testing corrections (**Supplementary Tables 22 and 23**).

## 12. Integration of additional phenotypic layers identifies additional loci implicated in post-challenge insulin action

Previous research has shown the utility of integrating information on associations with related traits when prioritising genetic loci associated with insulin resistance.<sup>23</sup> Here, we employed this approach to ISI and IFC with the aim to identify biologically relevant loci, with a post-challenge specific insulin resistance association pattern, which do not yet reach genome wide significance. This approach allowed us to identify additional loci of interest without sufficient sample size to detect these at a stringent genome wide significance threshold. To do this we prioritised variants that were suggestively associated with IFC and/or ISI ( $P < 5 \times 10^{-5}$ ). We next further filtered these variants based on their association with 2 h plasma glucose levels<sup>7</sup> ( $P < 5 \times 10^{-4}$ ) to prioritise those variants with additional evidence of being associated with post-challenge insulin resistance state. We additionally employed a second strategy where we filtered the variants suggestively associated with ISI and/or IFC based on their lack ( $P < 0.05$ ) of association with fasting insulin, to remove those likely associated with insulin resistance in the fasting state, implicating the liver. These variants were further filtered based on evidence of being eQTLs in skeletal muscle ( $P < 5 \times 10^{-4}$ ).<sup>24</sup> This allowed us to prioritise those variants with evidence of regulating expression in skeletal muscle, a tissue of high biological relevance to postprandial insulin action. Loci which met both prioritisation criteria were considered to have a post-challenge insulin resistance specific association signature, with further evidence of regulatory effects in skeletal muscle, the key tissue in post-prandial glucose uptake.

Through this strategy we prioritised 10 loci, including 6 loci that were not identified at genome wide significance for either IFC and ISI; **Supplementary Table 19**).

The *ERAP2* locus (rs1216570) was associated with higher IFC and 2 h glucose<sup>7</sup>, and lower ISI, suggestive of post-challenge insulin resistance (**Supplementary Figure 19a-b**. rs1216570 is an eQTL associated with higher *ERAP2* expression in skeletal muscle (GTEx v8 EUR: NES = 1.01, SE= 0.029,  $P = 4.15 \times 10^{-138}$ ). This eQTL in both muscle and adipose tissues, as well as 2hr glucose association has also been identified in other populations including in a cohort of African American ancestry<sup>25</sup>, as well as a cohort from Sweden.<sup>26</sup> A muscle *ERAP2* eQTL at this

locus has additionally identified in the Pima Indian population.<sup>27</sup> However, we identified limited evidence of a shared genetic signal for this eQTL signal (GTEx v8 EUR) and IFC (posterior probability of colocalisation = 0.32).

The lead variant at this locus, rs1216570, is additionally an eQTL in skeletal muscle for *LNPEP* (GTEx v8 EUR: NES = -0.13, SE = 0.026, P =  $3.92 \times 10^{-7}$ )<sup>22</sup> and there is strong evidence of colocalisation of this eQTL and IFC signal (posterior probability colocalisation = 0.85). *LNPEP* encodes IRAP, an insulin responsive amino peptidase that collocates with GLUT4 on GLUT4 secretory vesicles, and is well established to be involved in GLUT4 trafficking.<sup>28</sup> This demonstrates that this integrated approach identifies loci that are biologically relevant to post-challenge insulin resistance.

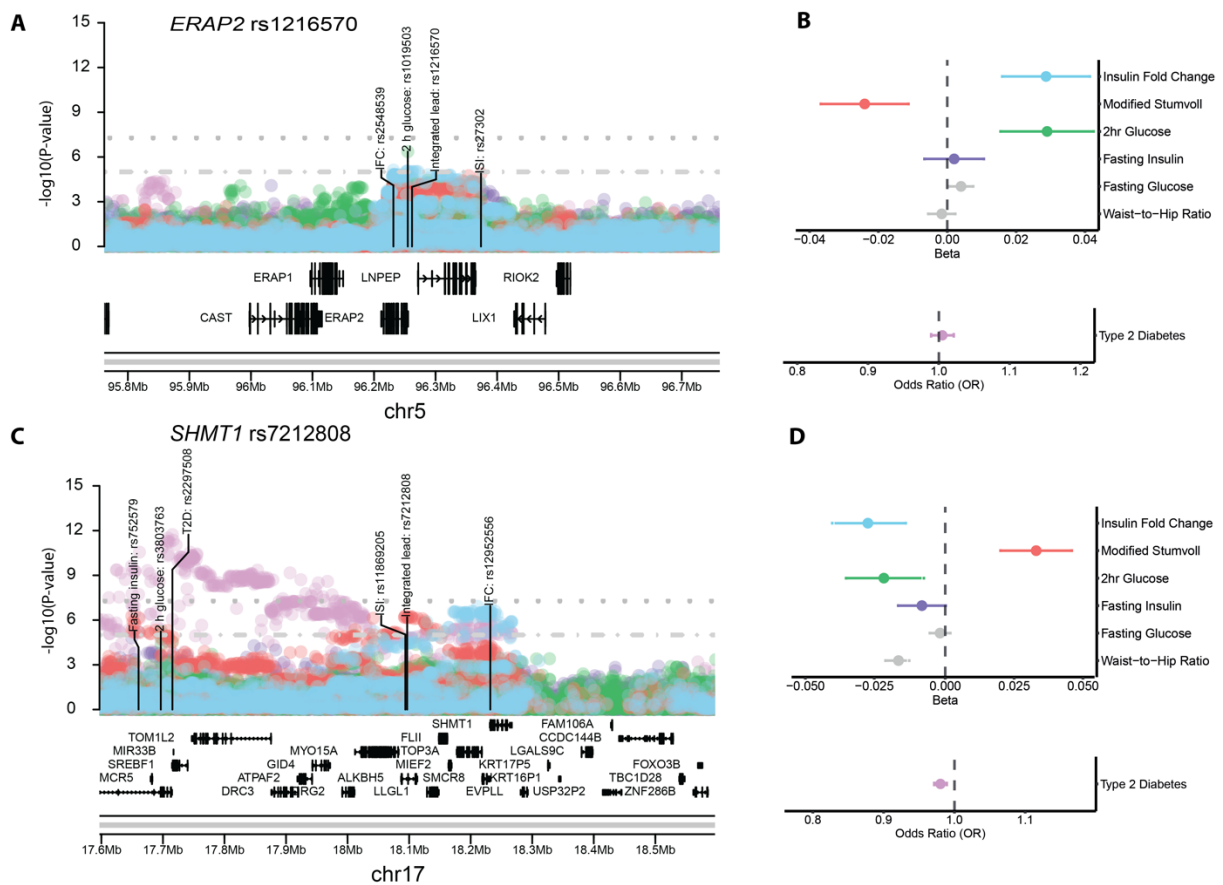
*SHMT1* (rs7212808) was associated with a favourable glycaemic signature (**Supplementary Figure 19c-d**) and is additionally an eQTL for increased *LLGL1* expression in skeletal muscle (GTEx v8 EUR: NES = 0.15, SE = 0.028, P =  $1.00 \times 10^{-7}$ ; posterior probability of colocalisation IFC and *LLGL1* eQTL = 0.91<sup>22</sup>). *LLGL1* is a Rab GTPase activating protein that interacts with Rab10, a key player in insulin-stimulated GLUT4 translocation.<sup>29</sup> Although *LLGL1* has not previously been implicated in post-challenge insulin response, it is plausible that this may play a role in regulating glucose uptake in the postprandial state.

All phenotypic signatures were additionally assessed using formal statistical colocalisation method *hyprcoloc*<sup>30</sup> (**Supplementary Figure 20**). We additionally tested for pairwise colocalisation of IFC and ISI signals with eQTLs in diabetes relevant tissues using *coloc*<sup>31</sup> (skeletal muscle, adipose, liver, and pancreatic islet; **see methods; Supplementary Table 20**). Further insight into tissue specificity of the genetic architecture of ISI and IFC was gained using LDSC-SEG, which leverages cell and tissue specific gene expression profiles to assess tissue specific enrichment of trait heritability (**see above; Supplementary Tables 21 to 25**).

By going below, the stringent genome wide significance level we identified evidence of loci directly implicating the biology of GLUT4 translocation in post-challenge insulin resistance comes from our findings at the *ERAP2* locus (rs1216570), suggestively associated (P =  $1.01 \times 10^{-$

<sup>5</sup>) with increased IFC and reduced ISI ( $P = 2.4 \times 10^{-4}$ ), and suggestively associated with 2 h glucose ( $P = 2.64 \times 10^{-5}$ ) but not fasting glycaemic measures. We identified evidence of a shared genetic signal for IFC and eQTL in skeletal muscle for *LNPEP* which encodes IRAP, the insulin responsive amino peptidase. IRAP is found on GLUT4 secretory vesicles and binds TBC1D4, to prevent translocation of GLUT4 to the plasma membrane in the fasting state (low insulin).<sup>32</sup> In line with previous reports that IRAP is a negative regulator of plasma membrane GLUT4<sup>28</sup>, depletion of *LNPEP* in 3T3-L1 adipocytes increased GLUT4 translocation to the cell surface. The lead variant at this locus is further in perfect LD ( $R^2=1$ ) in Europeans with a previously reported missense variant significantly associated with 2hr glucose (rs2549782)<sup>33</sup>. However, colocalisation of our lead variant rs1216570 with an *LNPEP* skeletal muscle eQTL, no evidence of an impact of *ERAP2* knockdown on GLUT4 trafficking in vitro, and in silico algorithms including SIFT<sup>34</sup>, PolyPhen2<sup>35</sup> and CADD<sup>36</sup> predicting the missense variant to be benign and to be unlikely to have an impact on the function of the encoding protein, suggest that *LNPEP* rather than *ERAP2* may be the causal gene at this locus.

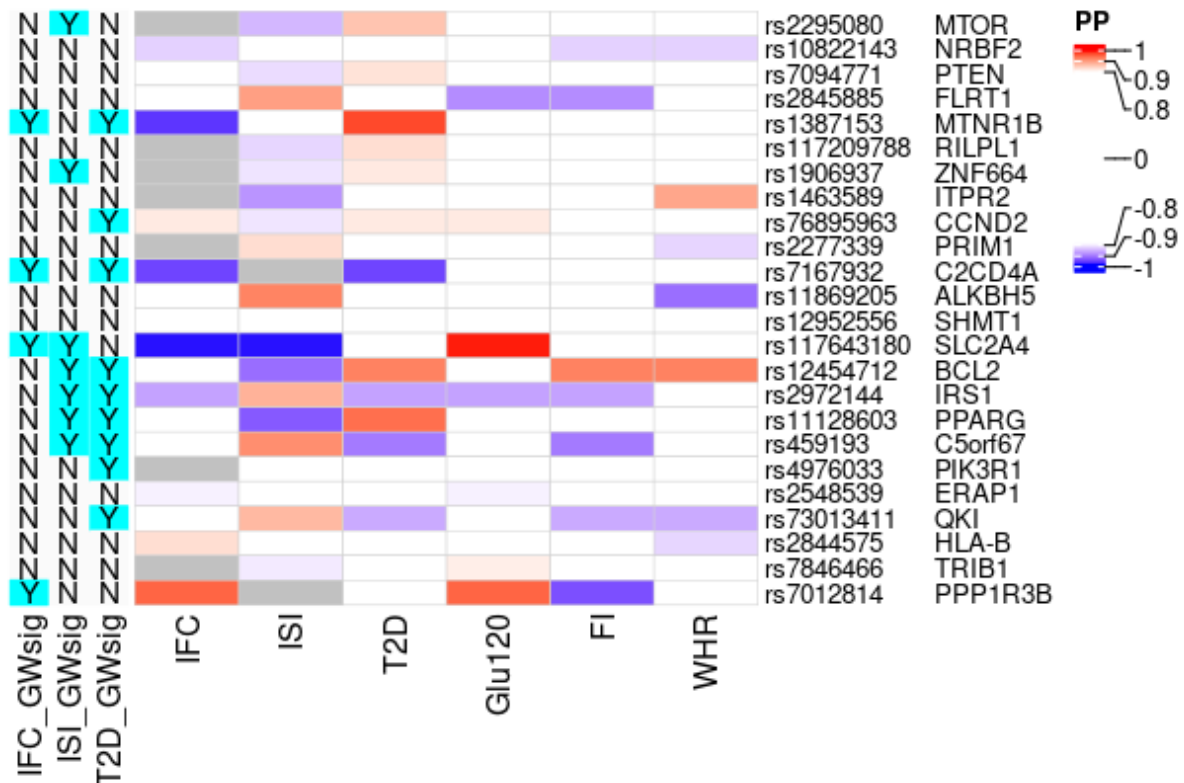
We further assessed potential tissue of action at individual associated loci by assessing colocalization of loci identified to be associated with IFC and Modified Stumvoll ISI in either the genetic discovery or integrated analyses with eQTLs in T2D relevant tissues – adipose, liver, pancreatic islets and skeletal muscle. We identified colocalization (posterior probability of colocalization  $> 0.7$ ) at 3 loci associated with IFC at *ERAP1/ERAP2*, *SHMT1* and *IRS1* with eQTLs identified in skeletal muscle and subcutaneous and visceral adipose tissues, key tissues in the post-prandial insulin response (**Supplementary Table 20**).



**Supplementary Figure 19: Integrated GWAS approach highlights additional loci including those implicating glucose transport.**

**A)** *ERAP2* locus: regional association plot for relevant traits. **B)** *ERAP2* locus: association of rs1216570 with relevant traits. **C)** *SHMT1* locus: regional association plot for relevant traits, with colours of points indicating traits of interest. **D)** *SHMT1* locus: association of rs7212808 with relevant traits. **A and C)** Regional association plot showing  $\pm 500$ kb around the lead IFC-associated variant, with points indicating the association and location of individual variants. Colours of points indicating traits of interest: insulin fold change (IFC) - blue, Modified Stumvoll (ISI) - red, and 2 h glucose<sup>7</sup> - green, fasting insulin<sup>7</sup> - purple, type 2 diabetes<sup>8</sup> - pink. Y-axis denotes the unadjusted  $-\log_{10}(p\text{-value})$  of association. The dashed lines and dotted line indicate genome-wide ( $P = 5 \times 10^{-8}$ ) and suggestive ( $P = 1 \times 10^{-5}$ ) significance threshold, respectively. X-axis denotes the genomic position. All association statistics are from BMI adjusted analyses in studies of European ancestry, except T2D which is unadjusted for BMI. The dashed lines indicate suggestive ( $P = 1 \times 10^{-5}$ ) significance threshold and dotted line indicates genome-wide significance ( $P = 5 \times 10^{-8}$ ), respectively. Labels indicate the lead variant for each trait, with location indicating position. Integrated lead is the lead variant defined for this locus using the integrated GWAS approach. **B and D)** Plot shows beta estimate or odds ratio of association for lead variant with insulin fold change (IFC) - blue, Modified Stumvoll (ISI) - red, and 2 h glucose<sup>7</sup> - green, fasting insulin<sup>7</sup> - purple, type 2 diabetes<sup>8</sup> - pink, all included in a) as well as fasting glucose<sup>7</sup> and waist-to-hip ratio<sup>9</sup> plotted in grey. All association statistics

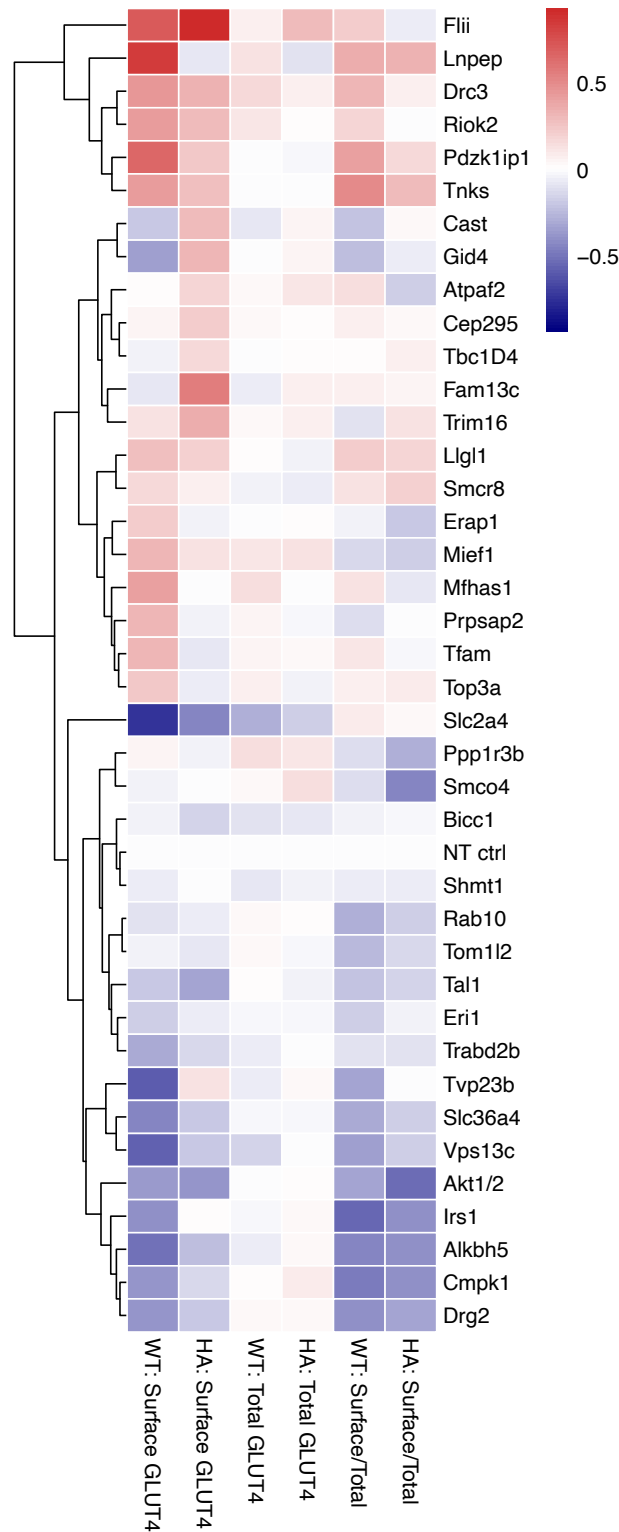
are from BMI adjusted analyses in studies of European ancestry. Error bars indicate 95% CI of the effect size. Sample size is outlined in **Methods**.



**Supplementary Figure 20: Assessment of colocalisation at loci associated with IFC or ISI with related traits**

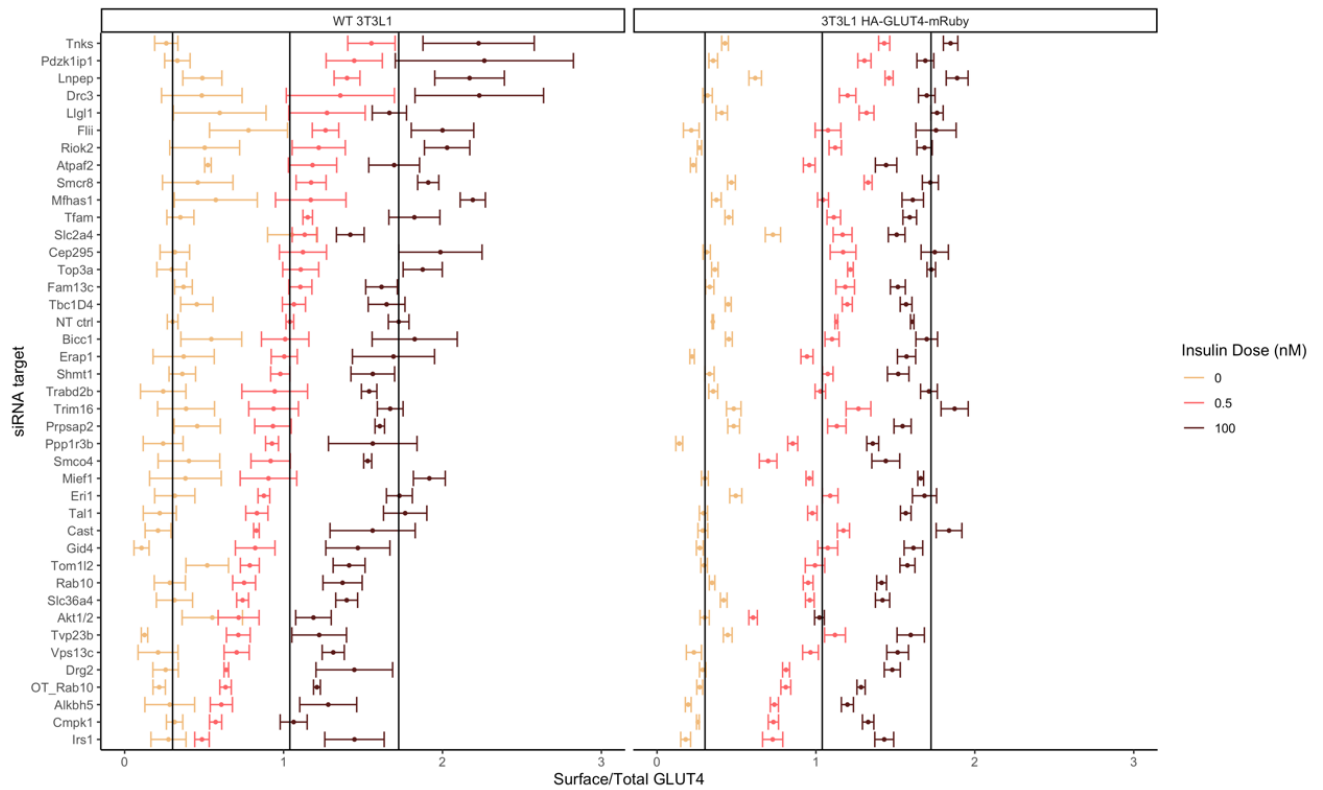
Fasting insulin (FI), 2hr glucose (Glu120) type 2 diabetes (T2D) and waist-to-hip ratio (WHR). All summary statistics used as input are from European ancestry individuals and adjusted for BMI, except for T2D which is unadjusted for BMI (Methods). GWsig denotes whether the IFC/ISI lead SNP at a locus (indicated by the rsid) meets genome-wide significance for the indicated trait (unadjusted  $P < 5e-8$ ). White denotes posterior probability  $< 0.8$ , with no evidence of colocalisation. Red positive beta for the indicated trait, the blue negative beta for the trait. Shade denotes the posterior probability of colocalisation between 0.8 and 1. Grey denotes that the trait was not included in hyprcoloc due to the locus (lead SNP +/- 500kb) not meeting the inclusion threshold of  $P < 5e-4$  for that trait. All effects are aligned to the IFC effect allele.





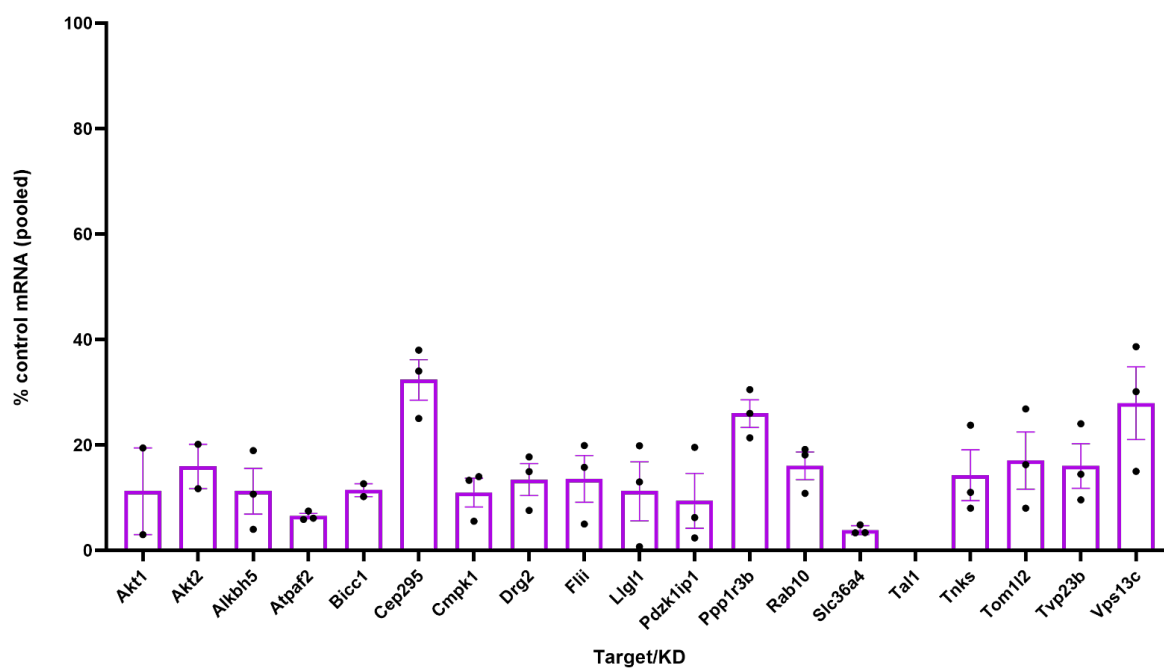
**Supplementary Figure 21: GLUT4 screen results for 3T3L1 WT and 3T3L1 HA-GLUT4-mRuby 3T3-L1 adipocytes for all targets**

All values are normalised to non-targeting control (NT) for the relevant measure/cell line combination. Targets with values lower than NT control are shown in blue and those with higher values in red. Clustering of targets was done using hierarchical clustering. Values are those measured at 0.5nM insulin. WT N = 3 biological replicates, with 2 technical replicates per N. HA – N = 5 biological replicates, with 2 technical replicates per N.



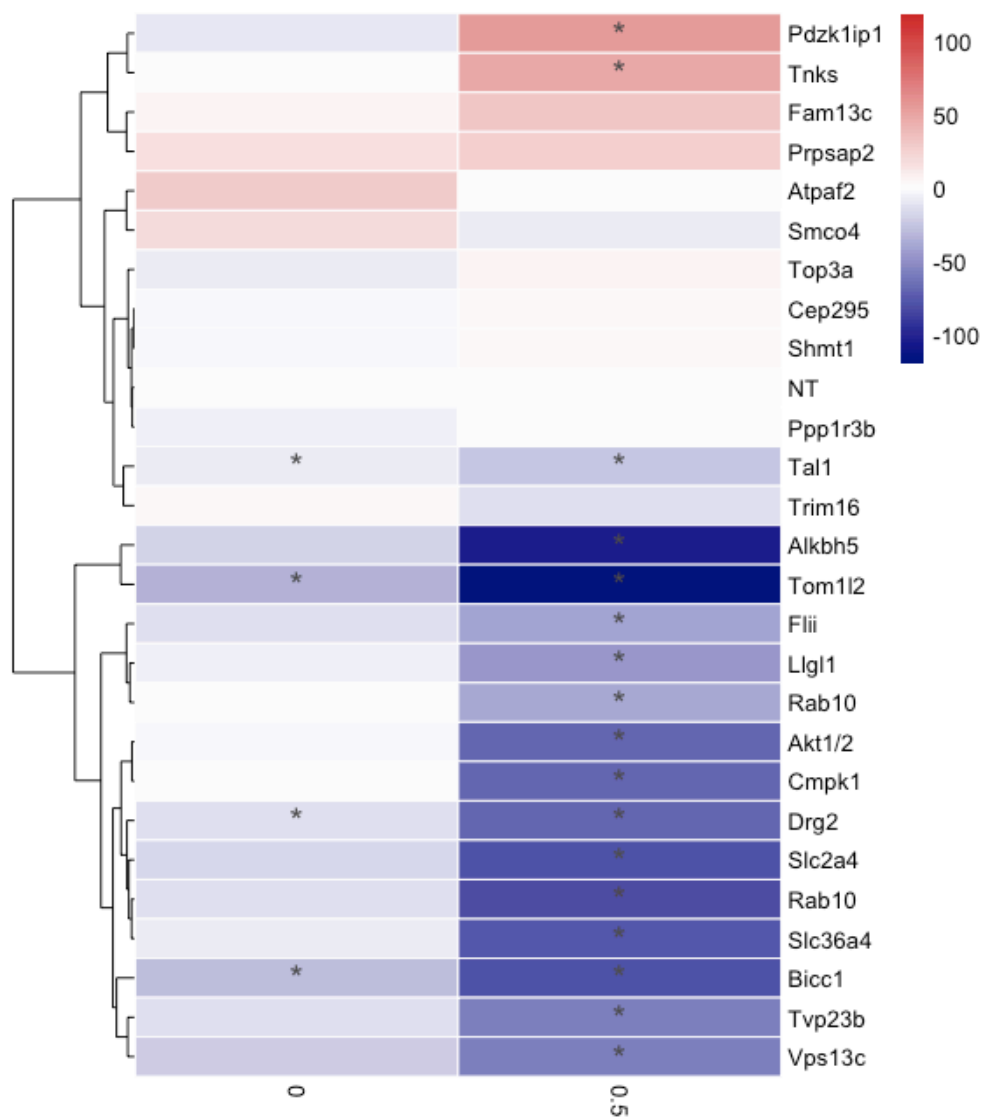
**Supplementary Figure 22: Overall screen surface/total GLUT4 comparison for HA-GLUT4-mRuby and WT 3T3-L1 adipocytes.**

Lines denote NT ctrl response at each dose line for WT 3T3L1 adipocytes. WT N = 3 biological replicates, with 2 technical replicates per N. HA-GLUT4-mRuby – N = 5 biological replicates, with 2 technical replicates per N. Insulin dose is indicated by colours in legend in nM. Error bar represents mean +/- SE.



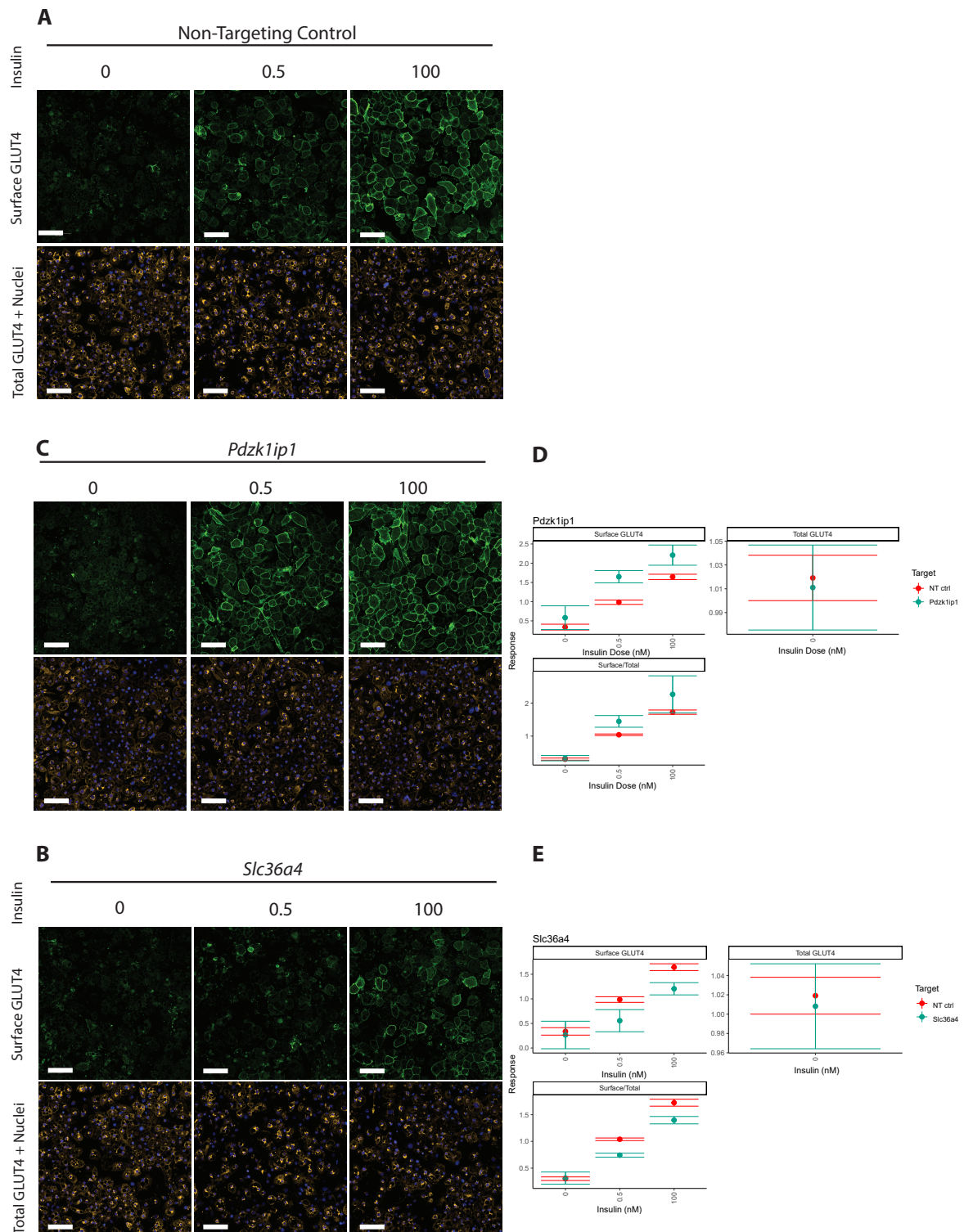
**Supplementary Figure 23: Average percentage of mRNA detected in knockdown samples compared to NT control.**

N = 2-3 independent biological replicates per gene, points represent independent samples. Error bar represents +/- standard error of mean.



**Supplementary Figure 24: Glucose Transport relative to NT control**

Values are normalised glucose response to non-targeting control (NT; 0) per dose of insulin (nM), and scaled to the absolute maximal response expressed as percentage. \* Indicates targets that are significantly different from the NT control at each dose (unadjusted  $P < 0.05$ ; two-tailed t-test).  $N = 3$  biological replicates per condition, except NT  $N = 6$ ; Akt1/2  $N = 6$ , Rab10  $N = 6$ , Tvp23b  $N = 4$ . All replicates were successful.



**Supplementary Figure 25: Visualisation and quantification of impact on parameters of interest of knockdown *Slc36a4* and *Pdk1ip1* in 3T3-L1 adipocytes.**

A) Representative microscopy images of non-targeting control (NT ctrl) at 0,0.5 and 100nM insulin stimulation. Surface GLUT4 in green, total GLUT4 in orange, and nuclei in blue. The scale bar on each image represents 100µm. B) Representative microscopy images of *Pdk1ip1* knockdown. As outlined in a). C) Representative microscopy images of *Slc36a4* knockdown. As outlined in a). D and E) quantification of the effect of knockdown on parameters of interest, normalised to background control. Response is relative to NT at a given insulin concentration, as indicated on X axis. Green – target, red- NT control. Error bars represent mean +/- SE.

## References for Supplementary Note:

1. Bycroft, C. *et al.* The UK Biobank resource with deep phenotyping and genomic data. *Nature* **562**, 203–209 (2018).
2. Lindsay, T. *et al.* Descriptive epidemiology of physical activity energy expenditure in UK adults (The Fenland study). *Int. J. Behav. Nutr. Phys. Act.* **16**, (2019).
3. Pietzner, M. *et al.* Genetic architecture of host proteins involved in SARS-CoV-2 infection. *Nat. Commun.* **2020 111 11**, 1–14 (2020).
4. Powell, R. *et al.* Development and validation of total and regional body composition prediction equations from anthropometry and single frequency segmental bioelectrical impedance with DEXA. *medRxiv* 2020.12.16.20248330 (2020) doi:10.1101/2020.12.16.20248330.
5. Lam, B. Y. H. *et al.* MC3R links nutritional state to childhood growth and the timing of puberty. *Nat.* **2021 5997885 599**, 436–441 (2021).
6. Bulik-Sullivan, B. *et al.* LD Score regression distinguishes confounding from polygenicity in genome-wide association studies. *Nat. Genet.* **2015 473 47**, 291–295 (2015).
7. Chen, J. *et al.* The trans-ancestral genomic architecture of glycemic traits. *Nat. Genet.* **2021 536 53**, 840–860 (2021).
8. Vujkovic, M. *et al.* Discovery of 318 new risk loci for type 2 diabetes and related vascular outcomes among 1.4 million participants in a multi-ancestry meta-analysis. *Nat. Genet.* **2020 527 52**, 680–691 (2020).
9. Pulit, S. L. *et al.* Meta-analysis of genome-wide association studies for body fat distribution in 694 649 individuals of European ancestry. *Hum. Mol. Genet.* **28**, 166–174 (2019).
10. Finucane, H. K. *et al.* Heritability enrichment of specifically expressed genes identifies disease-relevant tissues and cell types. *Nat. Genet.* **50**, 621 (2018).
11. Roadmap Epigenomics Consortium *et al.* Integrative analysis of 111 reference human epigenomes. *Nature* **518**, 317–329 (2015).
12. Dunham, I. *et al.* An integrated encyclopedia of DNA elements in the human genome. *Nature* **489**, 57–74 (2012).
13. Diaz-Vegas, A. *et al.* A high-content endogenous GLUT4 trafficking assay reveals new aspects of adipocyte biology. *Life Sci. alliance* **6**, (2023).
14. Willer, C. J., Li, Y. & Abecasis, G. R. METAL: Fast and efficient meta-analysis of genomewide association scans. *Bioinformatics* **26**, 2190–2191 (2010).
15. Wang, Q. *et al.* Rare variant contribution to human disease in 281,104 UK Biobank exomes. *Nat.* **2021 5977877 597**, 527–532 (2021).
16. American Diabetes Association. 2. Classification and diagnosis of diabetes: Standards of medical care in diabetes-2021. *Diabetes Care* **44**, S15–S33 (2021).
17. Spracklen, C. N. *et al.* Identification of type 2 diabetes loci in 433,540 East Asian individuals. *Nature* **582**, 240 (2020).
18. Williams, A. L. *et al.* Sequence variants in SLC16A11 are a common risk factor for type 2 diabetes in Mexico. *Nature* **506**, 97–101 (2014).
19. Karczewski, K. J. *et al.* The mutational constraint spectrum quantified from variation in 141,456 humans. *Nature* **581**, 434–443 (2020).
20. Auton, A. *et al.* A global reference for human genetic variation. *Nature* **526**, 68–74 (2015).
21. Machiela, M. J. & Chanock, S. J. LDlink: a web-based application for exploring

- population-specific haplotype structure and linking correlated alleles of possible functional variants. *Bioinformatics* **31**, 3555–3557 (2015).
22. Aguet, F. *et al.* The GTEx Consortium atlas of genetic regulatory effects across human tissues. *Science* (80-. ). **369**, 1318–1330 (2020).
  23. Lotta, L. A. *et al.* Integrative genomic analysis implicates limited peripheral adipose storage capacity in the pathogenesis of human insulin resistance. *Nat Genet* **49**, 17–26 (2017).
  24. Lonsdale, J. *et al.* The Genotype-Tissue Expression (GTEx) project. *Nature Genetics* vol. 45 580–585 (2013).
  25. Sajuthi, S. P. *et al.* Mapping adipose and muscle tissue expression quantitative trait loci in African Americans to identify genes for type 2 diabetes and obesity. *Hum. Genet.* **135**, 869–880 (2016).
  26. Keildson, S. *et al.* Expression of Phosphofructokinase in Skeletal Muscle Is Influenced by Genetic Variation and Associated With Insulin Sensitivity. *Diabetes* **63**, 1154–1165 (2014).
  27. Mason, C. C. *et al.* Bimodal distribution of RNA expression levels in human skeletal muscle tissue. *BMC Genomics* **12**, (2011).
  28. Jordens, I., Molle, D., Xiong, W., Keller, S. R. & McGraw, T. E. Insulin-regulated aminopeptidase is a key regulator of GLUT4 trafficking by controlling the sorting of GLUT4 from endosomes to specialized insulin-regulated vesicles. *Mol. Biol. Cell* **21**, 2034–2044 (2010).
  29. Sano, H. *et al.* Rab10, a target of the AS160 Rab GAP, is required for insulin-stimulated translocation of GLUT4 to the adipocyte plasma membrane. *Cell Metab.* **5**, 293–303 (2007).
  30. Foley, C. N. *et al.* A fast and efficient colocalization algorithm for identifying shared genetic risk factors across multiple traits. *Nat. Commun.* 2021 121 **12**, 1–18 (2021).
  31. Giambartolomei, C. *et al.* Bayesian Test for Colocalisation between Pairs of Genetic Association Studies Using Summary Statistics. *PLOS Genet.* **10**, e1004383 (2014).
  32. Rubio-Sastre, P., Scheer, F. A. J. L., Gómez-Abellán, P., Madrid, J. A. & Garaulet, M. Acute melatonin administration in humans impairs glucose tolerance in both the morning and evening. *Sleep* **37**, 1715-1719B (2014).
  33. Ng, N. H. J. *et al.* Tissue-Specific Alteration of Metabolic Pathways Influences Glycemic Regulation. *bioRxiv* **29**, 790618 (2019).
  34. Sim, N.-L. *et al.* SIFT web server: predicting effects of amino acid substitutions on proteins. *Nucleic Acids Res.* **40**, W452–W457 (2012).
  35. Adzhubei, I. A. *et al.* A method and server for predicting damaging missense mutations. *Nat. Methods* **7**, 248–249 (2010).
  36. Rentzsch, P., Witten, D., Cooper, G. M., Shendure, J. & Kircher, M. CADD: predicting the deleteriousness of variants throughout the human genome. *Nucleic Acids Res.* **47**, D886–D894 (2019).