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A low-frequency inactivating *AKT2* variant enriched in the Finnish population is associated with fasting insulin levels and type 2 diabetes risk

A full list of authors and affiliations appears at the end of the article.

These authors contributed equally to this work.

Abstract

To identify novel coding association signals and facilitate characterization of mechanisms influencing glycemic traits and type 2 diabetes risk, we analyzed 109,215 variants derived from exome array genotyping together with an additional 390,225 variants from exome sequence in up to 39,339 normoglycemic individuals from five ancestry groups. We identified a novel association between the coding variant (p.Pro50Thr) in *AKT2* and fasting insulin, a gene in which rare fully penetrant mutations are causal for monogenic glycemic disorders. The low-frequency allele is associated with a 12% increase in fasting plasma insulin (FI) levels. This variant is present at 1.1%

Corresponding authors. Prof. Anna L Gloyn, Oxford Centre for Diabetes Endocrinology & Metabolism, University of Oxford, Churchill Hospital, Headington, Oxford, OX3 7LE, United Kingdom, anna.gloyn@drl.ox.ac.uk. Prof. Cecilia M Lindgren, The Big Data Institute, Li Ka Shing Centre for Health Information and Discovery, The Wellcome Trust Centre for Human Genetics, University of Oxford, Roosevelt Drive, Oxford, OX3 7BN, United Kingdom, celi@well.ox.ac.uk.

&These authors jointly directed this research.

Author Contributions:

Sample Collection and Phenotyping: NG, A Mahajan, NPB, C Ladenvall, JB-J, NRR, NWR, RAS, APG, AUJ, CJG, CB, D Buck, GB, GJ, HMS, JRH, J Murphy, JMJ, J Trakalo, KSS, MM, MN, M Hollensted, RO, SG, ARW, ATH, HEA, AC, RAD, A Stan áková, AHR, A Metspalu, AJF, A-CS, A Käräjämäki, YAK, RA, A Swift, TA, BL, BG, BIF, B-GH, C Meisinger, CG, C Langenberg, D Pasko, D Aguilar, D Bowden, DH, EST, EC, C-YC, WYL, EM, SPF, FBH, G Atzmon, GWSr, DEH, HG, HK, HO, HATJr, TI, JSK, J Sehmi, J Lindstrom, J Kravic, JEC, CPJ, JEB, J Kriebel, JH, J Li, J Fadista, JCC, JCL, KRO, KSC, C-CK, LLB, J-YL, LK, DML, LH, L Milani, J Liu, L Liang, M Loh, MO-M, MW, MM-N, TM, MG, MR, MCYN, NDP, NN, LQ, NJW, NB, OM, OR, PJH, PWF, PN, A Peters, QQ, RM, S-TT, S Kumar, SKM, SPO'R, S Puppala, K Strauch, TMF, TK, TE, FT, BT, TVV, TYW, TAL, T Lauritzen, T Forsén, TIP, UA, VSF, WRS, YSC, ADM, ASFD, AL, BI, CNAP, FSC, CC, EI, FK, GLS, I Brandslund, J Tuomilehto, J Kuusisto, L Lannfelt, L Lind, LG, MEJ, MU, OP, RR, RNB, T Tuomi, TDS, TH, TJ, VS, GIB, JGW, JB, NJC, RD, KLM, M Laakso, CLH, APM, MB, D Altshuler, MIM

Replication and Expression Studies: T Tukiainen, AV, AAB, YW, A Palotie, AJ, JGE, OTR, S Koskinen, T Lehtimäki, JW, AYC, RAS, MOG, VS, JD, SR, JCF, JBM, ML, KLM

Data Production (Sequencing and Genotyping): XS, NG, A Mahajan, CF, NPB, C Hartl, C Ladenvall, JB-J, NRR, NWR, APG, AUJ, CJG, CB, D Buck, GB, GJ, HMS, JRH, J Murphy, JMJ, J Trakalo, KSS, MM, MN, M Hollensted, RO, PSC, SG, MOC, MD, EB, YF, MHdA, K Shakir, RP, T Fennell, TS, TW, TMS, K Stirrups, TM, PD, MB, MIM

Variant Calling & Panel Generation: MAR, KJG, HMK, GJ, BMN, GG, J Maguire, J Carey, JDS, JIG, S Purcell

Statistical Analysis: A Manning, HMH, JG, XS, T Tukiainen, P Fontanillas, NG, MAR, A Mahajan, AEL, P Cingolani, T Pers, J Flannick, CF, ERG, KJG, HKI, TMT, A Kumar, NPB, C Hartl, C Ladenvall, HMK, JB-J, YC, JRBP, LJS, C Ma, MvdB, L Moutsianas, NRR, RDP, TWB, TG, NWR, APG, AUJ, CJG, CB, D Buck, GB, GJ, HMS, JRH, J Murphy, JMJ, J Trakalo, KSS, MM, MN, M Hollensted, RO, SG, JBM, APM, MB, MIM, CML

Functional Studies: JG, SJ, ALG

Wrote the paper: A Manning, HMH, JG, XS, T Tukiainen, P Fontanillas, JCF, MB, MIM, ALG, CML

Study Design: XS, LJS, ATH, HEA, RAD, BG, EST, G Atzmon, JSK, CPJ, JCC, KSC, J-YL, DML, TM, TMF, TIP, YSC, C Hu, GRC, D Bharadwaj, PJD, D Prabhakaran, EZ, I Barroso, J Scott, J Chan, GM, MJD, M Sandhu, NT, PE, P Froguel, RCWM, RS, SBE, YYT, T Park, T Fingerlin, WJ, RMW, J Tuomilehto, LG, GIB, G Abecasis, JGW, JB, M Seielstad, NJC, RD, JD, IP, JCF, KLM, M Laakso, JBM, CLH, APM, MB, D Altshuler, MIM, ALG, CML

Study Supervision: GIB, G Abecasis, JGW, JB, M Seielstad, NJC, RD, JCF, KLM, JBM, CLH, APM, MB, D Altshuler, MIM, ALG, CML

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frequency in Finns but virtually absent in individuals from other ancestries. Carriers of the FI-increasing allele had increased 2-hour insulin values, decreased insulin sensitivity, and increased risk of type 2 diabetes (odds ratio=1.05). In cellular studies, the AKT2-Thr50 protein exhibited a partial loss of function. We extend the allelic spectrum for coding variants in *AKT2* associated with disorders of glucose homeostasis and demonstrate bidirectional effects of variants within the pleckstrin homology domain of *AKT2*.

The increasing prevalence of type 2 diabetes is a global health crisis, making it critical to promote development of more efficient strategies for prevention and treatment. Individuals with type 2 diabetes display both pancreatic beta-cell dysfunction and insulin resistance. Genetic studies of surrogate measures of these glycemic traits can identify variants that influence these central features of type 2 diabetes (2) highlighting potential pathways for therapeutic manipulation. Comprehensive surveys of the influence of common genetic variants on fasting plasma glucose (FG) and fasting plasma insulin (FI) have highlighted defects in pathways involved in glucose metabolism, and insulin processing, secretion, and action (3). Recent studies have identified type 2 diabetes-associated alleles that are common in one population but rare or absent in others (4–6). These associations were observed either due to an increase in frequency of older alleles based on population dynamics and demography (5), or the emergence of population-specific alleles (4; 6).

We set out to identify and characterize low-frequency allele (minor allele frequency [MAF]<5%) glycemic trait associations by meta-analysis of exome sequence and exome array genotype data in a multi-ancestry sample. We also performed *in vitro* functional studies of protein expression, localization and activity to understand the consequences of our novel findings.

Methods

Genetic association studies

Study Samples—The Genetics of Type 2 Diabetes (GoT2D) study and Type 2 Diabetes Genetic Exploration by Next-generation sequencing in multi-Ethnic Samples (T2D-GENES) study were initially designed to evaluate the contribution of coding variants to type 2 diabetes risk (7). We performed a discovery association analysis to find novel coding variants associated with fasting glycemic traits in 14 studies from GoT2D that contributed exome array information on 33,231 non-diabetic individuals of European ancestry. Further discovery analysis was performed with GoT2D and T2D-GENES studies with exome sequence data (average 80x coverage) in five ancestral groups comprised of 12,940 individuals (6,504 with type 2 diabetes, 6,436 without) with measured FG or FI levels available in 2,144 European, 508 South Asian, 1,104 East Asian, 844 Hispanic, and 508 African American non-diabetic individuals. We performed a replication analysis and an assessment of allele frequency distributions in 5,747 individuals from four Finnish cohorts: Cardiovascular Risk in Young Finns Study (YFS) (8), Helsinki Birth Cohort (HBCS) (9), Health 2000 GenMets Study (GenMets) (10), and National FINRISK Study 1997 and 2002 (FR) (11). We also assessed the allele frequencies of novel findings in 46,658 individuals from CHARGE studies with available exome array data (12), although none of the studies

passed our QC filter of a minor allele count greater than 5 for inclusion in our replication analysis. See Supplementary Table 1 for study details, sample characteristics, ascertainment criteria, and detailed genotype calling and quality control procedures for each cohort. The relevant institutional review boards, conducted according to the Declaration of Helsinki, approved all human research and all participants provided written informed consent. A detailed description of ethical permissions is provided in the Supplementary Materials.

Phenotypes—For the discovery and replication analysis, we excluded individuals from the analysis if they had a diagnosis of type 2 diabetes, were currently receiving oral or injected diabetes treatment, had FG measures ≥ 7 mmol/L, had 2-hour post-load glucose (2hrG) measures ≥ 11.1 mmol/L, or had HbA1c measures $\geq 6.5\%$ (48 mmol/mol). Additional exclusions occurring at the study level included pregnancy, non-fasting at time of exam, type 1 diabetes, or impaired glucose tolerance. See Supplementary Table 1A for details. Within each study, we adjusted FG and log transformed FI levels for age, sex, body mass index (BMI), and additional study specific covariates. We applied rank-based inverse-normal transformations to study- or ancestry-specific residuals to obtain satisfactory asymptotic properties of the exome-wide association tests.

We tested for genetic associations with type 2 diabetes, hypertension, and other related quantitative traits in the Finnish discovery and replication cohorts. We analyzed lipid levels (total cholesterol, high-density lipoprotein cholesterol (HDL), low-density lipoprotein cholesterol (LDL), and triglycerides (TG)), blood pressure (systolic (SBP) and diastolic (DBP) blood pressures and hypertension (HTN)), height, BMI, central adiposity measures (waist-to-hip ratio (WHR), waist circumference, hip circumference), adiponectin level, 2-hour insulin level, and Matsuda index, which is known to correlate with whole-body insulin sensitivity as measured by the hyperinsulinemic euglycemic clamp ($r=0.7$, $P<1.0\times 10^{-4}$) (13). For quantitative traits and HTN, we adjusted for age, sex, BMI (for glycemic, blood pressure, and central adiposity traits), stratified by type 2 diabetes status and sex (for central adiposity measures) within study. We adjusted LDL and total cholesterol for use of lipid-lowering medication, by dividing total cholesterol by 0.8 if on lipid-lowering medication, prior to calculating LDL using the Friedewald equation (14). SBP and DBP were adjusted for use of blood pressure-lowering medication by adding 15 mmHg to SBP and 10 mmHg to DBP measurements if an individual reported taking blood pressure-lowering medication (15). The Matsuda Index was log transformed and analyzed in non-diabetic individuals only. After adjusting for covariates, traits were inverse-normalized within strata. In addition to studying these metabolic outcomes, we used international classification of diseases (ICD) codes to query electronic medical records in the METSIM and FINRISK 1997 and 2002 cohorts (in all individuals regardless of type 2 diabetes status) and categorized affection status for lipodystrophy, polycystic ovary disease, and ovarian or breast cancer.

Statistical Analysis

Discovery Analysis: We performed association analyses within each study for the exome array data sets and within ancestry for the exome sequence data sets. We used linear mixed models implemented in EMMAX (16) to account for relatedness. Within each study/ancestry, we required variants to have a minor allele count (MAC) greater than or equal to

five alleles for single variant association tests. We meta-analyzed the single variant results from the (European-ancestry) exome array studies using the inverse variance meta-analysis approach implemented in METAL (17) and combined these with the European ancestry exome sequence results. Then, we meta-analyzed summary statistics across ancestries. We used $P < 5 \times 10^{-7}$ as exome-wide statistical significance thresholds for the single variant tests (18). We used the binomial distribution to assess enrichment of previously reported associations with FG or FI by calculating a P -value for the number of non-significant variants with consistent direction of effects.

Gene based association analysis: We performed gene-based association tests using variants with $MAF < 1\%$ (including rare variants with $MAC \geq 5$), annotating and aggregating variants based on predicted deleteriousness using previously described methods (7). Briefly, we defined four different variant groupings: “PTV-only”, containing only variants predicted to severely impair protein function, “PTV+missense”, containing PTV and NS variants with $MAF < 1\%$, “PTV+NS_{strict}” composed of PTV and NS variants predicted damaging by five algorithms (SIFT, LRT, MutationTaster, polyphen2 HDIV, and polyphen2 HVAR), and “PTV+NS_{broad}” composed of PTV and NS variants with $MAF < 1\%$ and predicted damaging by at least one prediction algorithm above. We used the sequence kernel association test (SKAT) (19) and a frequency-weighted burden test to conduct exome array meta-analyses in an unrelated subset of individuals using RareMETAL (20). We conducted exome sequence gene-based analyses within ancestry using a linear mixed model to account for relatedness and combined results across ancestries with MetaSKAT (21), which accounts for heterogeneous effects. We further combined gene-based results from exome array and exome sequences using Stouffer’s method with equal weights. For gene-based tests, we considered $P < 2.5 \times 10^{-6}$ as exome-wide significant, corresponding to Bonferroni correction for 20,000 genes in the genome (18).

Replication Analysis: The *AKT2* p.Pro50Thr variant was observed at sufficient frequency in the independent Finnish cohorts to perform single-variant association test of association with FI. We tested association in SNPTEST (22) (v.2.4.0) in each study with the same additive linear model used in the discovery analysis. Covariate adjustments for FI levels were sex, age, and ten principal components (PCs), and models were run with and without adjustment for BMI.

Estimate of effect on raw FI level and variance explained: To characterize the association between *AKT2* p.Pro50Thr and FI, we examined full regression models with raw FI in three studies (FUSION, METSIM, and YFS). We estimated the raw effect on log-transformed FI levels with a fixed-effects meta-analysis. The variance in log-transformed FI explained by *AKT2* p.Pro50Thr was estimated by a weighted average of the narrow-sense heritability of *AKT2* p.Pro50Thr seen in these three studies.

Population genetics and constraint: We used the Exome Aggregation Consortium (ExAC) for constraint metrics and allele frequencies (23). We obtained sequence alignments for AKT proteins and mRNAs in 100 vertebrates from the UCSC Genome Browser (24), used

Shannon's entropy (normalized $K=21$) as a conservation score (25) and plotted the sequence logos in R using the RWebLogo library (26).

Associations with other traits: We conducted association tests for traits other than FI and FG within studies for both discovery studies as well as the independent Finnish studies used for replication. P -values for type 2 diabetes and HTN came from EMMAX (16) or the Wald test from logistic regression (Finnish replication data sets) and meta-analyzed using an N weighted meta-analysis (17). Odds ratios (OR) were obtained from logistic regression adjusting for age, sex, with and without BMI, and PCs and meta-analyzed using an inverse variance meta-analysis.

Trait distributions and phenotype clustering: We examined distributions of traits among *AKT2* missense allele carriers (p.Pro50Thr, p.Arg208Lys, and p.Arg467Trp) in the T2D-GENES exome sequencing data set. We used non-parametric rank based methods (kruskal.wallis and permKS functions in R) on both the inverse-normalized covariate-adjusted traits used in the genetic association studies and normalized raw trait values (scale function in R). We clustered *AKT2* missense allele carriers on scaled trait values (pheatmap function in R).

***In vitro* functional studies**

Plasmids and cell lines—The generation of the *AKT2* allelic series was initiated by the production of pDONR223-AKT2 through PCR of the human *AKT2* open reading frame with the integration of terminal attR sites using primers (see below). HeLa, HuH7, and 293T cells were obtained at The Broad Institute and maintained in 10% FBS DMEM, 100U/ml penicillin and 100 μ g/ml streptomycin, and documented mycoplasma-free. HeLa and HuH7 cells were starved for 18 hours and stimulated for 15 minutes with 100nM insulin for activation analyses.

Primers for functional work—The generation of the *AKT2* allelic series was initiated by the production of pDONR223-AKT2 through PCR of the human *AKT2* open reading frame with the integration of terminal attR sites using primers FWD: 5'-GGGGACAAGTTTGTACAAAAAAGTTGGCACCATGAATGAGGTGTCTGTCATC-3' REV: 5'-GGGGACCACTTTGTACAAGAAAGTTGGCAACTCGCGGATGCTG-3', and subsequent Gateway BP reaction into pDONR223 obtained from The Broad Institute Genetics Perturbation Platform. Site-directed mutagenesis was then performed to generate AKT2.E17K (AKT2.Lys17), AKT2.P50T (AKT2.Thr50), AKT2.R208K (AKT2.Lys208), AKT2.R274H (AKT2.His274), AKT2.R467W (AKT2.Trp467) with the following primers: AKT2.E17K: FWD: 5'-GGCTCCACAAGCGTGGTAAATACATCAAGACCTGG-3' REV: 5'-CCAGGTCTTGATGTATTTACCACGCTTGTGGAGCC-3'; AKT2.P50T: FWD: 5'-AGGCCCTGATCAGACTCTAACCCCTTAAAC-3' REV: 5'-GTTAAGGGGGTTAGAGTCTGATCAGGGGCCT-3'; AKT2.R208K: FWD: 5'-GTCCCTCCAGAACACCAAGCACCCGTTCC-3' REV: 5'-GGAACGGGTGCTTGGTGTCTGGAGGAC-3'; AKT2.R274H: FWD: 5'-GGGACGTGGTATACCACGACATCAAGCTGGA-3' REV: 5'-TCCAGCTTGATGTCGTGGTATACCACGTCCTCC-3'; AKT2.R467W: FWD: 5'-

GGAGCTGGACCAGTGGACCCACTTCCC -3' REV: 5'-
GGGAAGTGGGTCCACTGGTCCAGCTCC -3'. C-terminal, V5-tagged lentiviral pLX304-AKT2.E17K, pLX304-AKT2.P50T, pLX304-AKT2.R208K, pLX304-AKT2.R274H, and pLX304-AKT2.R467W were each generated by subsequent Gateway LR reactions with pDONR223-AKT2.E17K, pDONR223-AKT2.P50T, pDONR223-AKT2.R208K, pDONR223-AKT2.R274H, and pDONR223-AKT2.R467W, respectively, and pLX304 obtained from The Broad Institute Genetics Perturbation Platform. Control plasmid pLX304-empty vector was additionally acquired from The Broad Institute Genetics Perturbation Platform.

Antibodies—Anti-Akt (#4685), anti-phospho-Akt S473 (#4060), anti-phospho-Akt T308 (#9275), anti- β Actin (#4970), anti-GSK3 β (#9315), anti-phospho-GSK3 β (#9336), anti-GST (#2625), and anti-V5 (#13202) were purchased from Cell Signaling Technologies (product numbers listed for each). Horseradish peroxidase-conjugated anti-rabbit and anti-mouse immunoglobulin G (IgG) antibodies were purchased from Millipore.

3D modeling—The 3D structure of AKT2 with the full allelic series was predicted using IntFOLD (27) and visualized in PyMOL (28).

In vitro kinase assays—We isolated V5-AKT2, V5-AKT2.Lys17, V5-AKT2.Thr50, V5-AKT2.Lys208, V5-AKT2.His274, and V5-AKT2.Trp467 variants from lentivirally infected and 5 μ g/mL blasticidin selected HeLa cell lysate with V5 agarose beads (SIGMA) and incubated with 150ng GST-GSK3 β substrate peptide (Cell Signaling Technologies) and 250mM cold ATP in kinase assay buffer (Cell Signaling Technologies) for 35 minutes at 30°C.

Proliferation assay—We cultured lentiviral pLX304 V5-AKT2 variants and control empty vector infected and 5 μ g/mL blasticidin selected HuH7 cells in 24 well plate for 72 hours in 10% FBS /phenol red-free DMEM for 72 hours. We added WST-1 (Takara Clontech) to each well at the manufacture recommended 1:10 ratio and incubated for 4 hours at 37°C prior to absorbance measurement at 450nm with BioTek Synergy H4 plate reader.

Immunoblots—We washed cells with phosphate buffered saline and lysed in EBC buffer (120mM NaCl, 50mM TRIS-HCl (pH7.4), 50nM calyculin, cOmplete protease inhibitor cocktail (Roche), 20mM sodium fluoride, 1mM sodium pyrophosphate, 2mM ethylene glycol tetraacetic acid, 2mM ethylenediaminetetraacetic acid, and 0.5% NP-40) for 20 minutes on ice. To preclear cell lysates, we centrifuged at 12,700 rpm at 4°C for 15 minutes. We measured protein concentration with Pierce BCA protein assay kit using a BioTek Synergy H4 plate reader. We resolved lysates on BioRad any kD mini-PROTEAN TGX polyacrylamide gels by SDS-PAGE and transferred by electrophoresis to nitrocellulose membrane (Life Technologies) at 100V for 70 minutes. We blocked membranes in 5% nonfat dry milk/ TBST (10mM Tris-HCl, 150mM NaCl, 0.2% Tween 20) buffer pH 7.6 for 30 minutes. We incubated blots with indicated antibody overnight at 4°C. The membrane was then washed in TBST, three times at 15 minute intervals, before 1 hour secondary horseradish peroxidase-conjugated antibody incubation at room temperature. We again

washed nitrocellulose membranes in TBST, three times for 15 minutes, prior to enhanced chemiluminescent substrate detection (Pierce).

Statistical analysis—The quantified results of the *in vitro* kinase and proliferation assays were normalized to internal control values for each replicate. We used generalized linear models of the quantified assay results to assess effects of variants within and across replicate rounds, allowing for interaction by replicate. The graphical representation was produced using functions in the effects (v 3.0-3) package in R.

Gene Expression Studies

Study samples—*GTEX*: We compared the expression pattern of *AKT2* to the two other members of the *AKT* gene family, *AKT1* and *AKT3*, using multi-tissue RNA sequencing (RNA-seq) data from the pilot phase of the GTEX project (dbGaP accession number: phs000424.v3.p1) in 44 tissues with data from more than one individual. Detailed procedures for sample collection, RNA extraction, RNA-seq, and gene and transcript quantifications have been previously described (29). *EuroBATs*: Samples from photo protected subcutaneous adipose tissue from 766 twins were extracted (130 unrelated individuals, 131 monozygotic and 187 dizygotic twin pairs) and processed as previously described (30; 31). *METSIM*: Subcutaneous fat biopsy samples were obtained from a sample of 770 participants from the METSIM study and processed as previously described (32).

Phenotypes—We studied the association of age, body mass index (BMI) and fasting insulin levels with gene expression levels and with expression-associated SNPs (eQTLs) in the *AKT2* region. Age and sex were available for the GTEX study samples. In addition to age and BMI, fasting insulin level was measured at the same time point as the fat biopsies in the EuroBATs sample data, following a previously described protocol (33). Baseline age, BMI and fasting insulin levels were used for the METSIM study participants (34)

Statistical analysis—The comparison of expression levels of *AKT2* versus *AKT1*, and *AKT2* versus *AKT3* was performed using log₂-transformed reads per kilobase per million mapped reads (RPKM_s). The percent increase in *AKT2* expression was calculated with the following formula: $2^{\Delta \log\text{-fold-change}} (AKT2 \text{ vs } AKT1)$. We studied BMI, age, and fasting insulin (not available in GTEX data) associations with *AKT2* expression using linear mixed models as implemented in the lme4 package in R. The gene expression RPKM values were inverse variance rank normalized for these analyses. Covariates included study-specific fixed and random effects (see Supplementary Note 4 for additional details on each cohort), using sex, BMI and age as additional fixed effects as appropriate. The expression quantitative trait loci (eQTL) analysis was performed on single nucleotide polymorphisms (SNPs) within a 1 Mb of *AKT2* using linear mixed models to assess the association of the SNPs with the inverse normalized RPKM expression values.

Results

Genetic association studies

We tested the association of FI and FG with 390,225 variants from exome sequence data (GoT2D and T2D-GENES studies) and 109,215 variants derived from exome array genotyping (GoT2D studies) (7) (individual study $\lambda_{GC} < 1.06$; Supplementary Figure S1). We examined variants that had been previously associated with FG and FI (3; 18). Of 28 FG and 14 FI loci with the reported SNPs or close proxies in our data set, 13 FG and four FI showed directionally consistent significant associations. Among the remaining GWAS loci not significant in our data, we observed directionally consistent associations in 14/15 FG and 9/10 FI loci ($P_{\text{enrichment}} = 5 \times 10^{-4}$ for FG and 0.01 for FI) (Supplementary Note 1; Supplementary Table 2).

In addition, we identified a novel significant single variant association between rs184042322 and FI (MAF=1.2%, $P=1.2 \times 10^{-7}$), a coding variant in *AKT2* (*V-AKT Murine Thymoma Viral Oncogene Homolog 2*) where amino acid Pro50 is substituted with a threonine (NP_001617.1:p.Pro50Thr) (Figure 1; Supplementary Figure S1). The same allele drove a significant FI signal for *AKT2* in gene-based analysis ($P=6.1 \times 10^{-7}$), in which we discovered two additional significant gene-based associations between *GIMAP8* and FG ($P_{\text{PTV}}=2.3 \times 10^{-6}$), and between *NDUFAF1* and FI ($P_{\text{PTV+NSBroad}}=9.2 \times 10^{-7}$) (Supplementary Figure S2; Supplementary Table 2D).

In an effort to replicate the single variant association of *AKT2* Pro50Thr with FI, we aggregated the allele frequency estimates of *AKT2* Pro50Thr in our data with data from the CHARGE consortium and the four Finnish studies. In ExAC, rs184042322 is multi-allelic (p.Pro50Thr and p.Pro50Ala) but Pro50Ala is observed only twice in the Latino population sample and not seen in our exome sequencing data, which includes 1,021 individuals of Hispanic ancestry. *AKT2* Pro50Thr was observed at a much higher frequency in Finnish individuals (MAF=1.1%) than other European (MAF=0.2%), African American (MAF=0.01%), Asian (MAF<0.01%), or Hispanic (MAF<0.01%) individuals (Figure 1). We replicated the association between FI and *AKT2* Pro50Thr by meta-analysis of the association in the four Finnish studies ($P=5.4 \times 10^{-4}$, N=5,747) with the discovery studies ($P_{\text{combined}}=9.98 \times 10^{-10}$, N=25,316). We observed no evidence of effect-size heterogeneity between studies ($P_{\text{Heterogeneity}}=0.76$). The minor T allele was associated with a 12% (95% CI=7%-18%) increase in FI levels in the discovery and replication studies, a per allele effect of 10.4pmol/L (95% CI=6.6-14.3pmol/L).

The serine/threonine protein kinases AKT1, AKT2, and AKT3 are conserved across all vertebrates (Figure 2). Pro50 and the seven preceding residues in the pleckstrin homology (PH) domain appear to be specific for the AKT2 isoform. Population genetic studies show a strong intolerance to missense and loss of function variation in *AKT2* (Supplementary Note 2; Supplementary Figure S3; Supplementary Figure S4; Supplementary Table 3). Notably, in ExAC data, *AKT2* contains fewer missense variants than expected (the missense constraint metric, $Z=3.5$, is in the 94th percentile of all genes) and extreme constraint against loss-of-function (LoF) variation (estimated probability of being LoF intolerant (pLI)=1).

AKT2 is a primary transducer of phosphoinositide 3-kinase (PI3K) signaling downstream of the insulin receptor and is responsible for mediating the physiological effects of insulin in tissues including liver, skeletal muscle, and adipose. *Akt2* null mice are characterized by hyperglycemia and hyperinsulinemia, and some develop diabetes (35; 36). In humans, highly penetrant rare alleles in *AKT2* cause familial partial lipodystrophy and hypoinsulinemic hypoglycemia with hemihypertrophy (Glu17Lys) (37; 38) and a syndrome featuring severe insulin resistance, hyperinsulinemia, and diabetes mellitus (Arg274His) (39). Additional rare alleles have been observed in individuals with severe insulin resistance (Arg208Lys and Arg467Trp) but no variant has been associated with glycemic traits at the population level (40).

Given the spectrum of diseases and traits associated with *AKT2* (41), we hypothesized that *AKT2* Pro50Thr would be associated with features of metabolic syndrome or lipodystrophy. In quantitative trait analysis in the initial discovery and replication cohorts, we did observe a constellation of features indicative of a milder 'lipodystrophy-like phenotype' associated with the rare allele: associations with increased 2-hour insulin values (effect=0.2 SD of log-transformed 2-hour insulin, 95% CI=0.1-0.4; $P=7.9\times 10^{-8}$, N=14,150), lower insulin sensitivity (effect=-0.3 SD of the log-transformed Matsuda index, 95% CI=-0.5 to -0.2, $P=1.2\times 10^{-6}$, N=8,566), and increased risk of type 2 diabetes (odds ratio (OR)=1.05 95% CI=1.0-1.1, $P=8.1\times 10^{-5}$; 9,783 type 2 diabetes cases; 22,662 controls), with no effects on fasting glucose, postprandial glucose, or fasting lipid levels ($P > 0.01$; Supplementary Table 4). In the T2D-GENES exome sequencing data where FG and FI levels were available in diabetic individuals, we observed one individual who was homozygous for the P50T allele with FI and FG levels in the 99.8th and 98.8th percentiles, respectively. There was a significant difference in trait distributions by P50T genotype (FI $P=0.002$; FG $P=0.02$; Supplementary Figure S5; Supplementary Table 4). Next, we used electronic health records available in the Finnish METSIM and FINRISK cohorts to characterize the impact of *AKT2* Pro50Thr on disease risk. We found no evidence for association with any cancer, polycystic ovary disease, or acanthosis nigricans (Supplementary Table 5); however, these tests are underpowered due to the low number of cases and potential for misclassification. Nor did we find evidence for enrichment of low-frequency associations in any *AKT2* related pathways or genes implicated in monogenic forms of glycemic disease (Supplementary Note 3; Supplementary Table 6; Supplementary Table 7; Supplementary Figure S6; Supplementary Figure S7).

***In vitro* functional studies**

To understand the functional consequences of the *AKT2* Pro50Thr variant on the protein, we investigated protein expression, activation, kinase activity, and downstream effector phosphorylation.

First, we used *in silico* classifiers that predict potential functional consequences of alleles on protein function. Two of the five classifiers predicted *AKT2* Pro50Thr to be deleterious (Supplementary Table 3). Second, we used 3D models of *AKT2* viewed in the PyMol software, which predicted that the Pro50Thr variant causes a change in the conformations of the lipid binding PH domain (Figure 3, Supplementary Figure S8). We hypothesized that the

variant protein is inefficiently recruited to the plasma membrane thereby impacting AKT2 phosphorylation and downstream activity.

To assess the molecular and cellular consequence of the *AKT2* Thr50 variant on protein function, we performed a comparative analysis of AKT2-Thr50 with inactivating and activating alleles implicated in monogenic disorders of insulin signaling. Analysis of AKT2-Thr50 expression showed that while AKT2 protein levels remained unchanged, there was a partial loss of AKT2-Thr50 phosphorylation at its activation sites (Thr308 and Ser473) in HeLa cells, suggesting impaired AKT2 signaling (Figure 3; Supplementary Figure S9). Similar effects were observed in human liver derived HuH7 cells (Supplementary Figure S10). AKT2-Thr50 also showed a reduced ability to phosphorylate its downstream target glycogen synthase kinase 3 beta (GSK3 β). These defects in AKT2-Thr50 activity were confirmed through an *in vitro* kinase assay ($P < 0.01$) (Figure 3). AKT2-Thr50 showed a similar decrease in kinase function to the lipodystrophy-causing AKT2-His274 variant. Using a four-hour time course analysis of AKT2 activity, we verified a reduction in both maximally phosphorylated Thr308 and Ser473 in AKT2-Thr50 (Supplementary Figure S11). To understand how this loss of activity could manifest as a defect in a known cellular function of AKT2 (42), we determined the impact of AKT2-Thr50 on cell proliferation in HuH7 cells. While the addition of AKT2 stimulated hepatocyte proliferation, the response to AKT2-Thr50 was reduced (effect = -1.2, $P < 1.0 \times 10^{-3}$) (Figure 3C; Supplementary Figure S12).

Gene expression studies

We queried RNA sequencing data from the Genotype Tissue Expression (GTEx) Project and found that, in agreement with previous studies (43), *AKT2* is highly and ubiquitously expressed across all tissues (44 tissue types, 3-156 individuals/tissue). Notably the *AKT2* Pro50Thr containing exon is expressed in all tissues and individuals (Supplementary Figure S13), suggesting that the PH domain is important to AKT2 function (44). Of the three *AKT* homologs, *AKT2* had 1.4-fold higher expression in skeletal muscle than *AKT1* ($P = 1.5 \times 10^{-19}$) and 11-fold higher expression than *AKT3* ($P = 7.8 \times 10^{-91}$). Skeletal muscle was the only tested tissue displaying such pronounced *AKT2* enrichment (Figure 2; Supplementary Note 4; Supplementary Figure S14; Supplementary Table 8).

Motivated by the age-related loss of adipose tissue in *Akt2* null mice (35; 36) and the growth and lipodystrophy phenotypes in carriers of fully-penetrant alleles (37–40), we examined associations of expression levels of *AKT2* with BMI, FI, and age in the three adipose tissue data sets (Supplementary Table 9). We found an association between lower BMI levels and higher *AKT2* expression in two cohorts (EuroBATS effect = -0.07 SD, $P = 6.1 \times 10^{-28}$; METSIM effect = -0.06 SD, $P = 8.1 \times 10^{-8}$) and also observed that higher *AKT2* expression was associated with lower log-transformed FI (EuroBATS, effect = -0.04 SD, $P = 1.1 \times 10^{-3}$, METSIM, effect = -0.4 SD, $P = 3.3 \times 10^{-11}$). We next tested for gene expression quantitative trait loci (eQTL) and found an eQTL in the 5'UTR of *AKT2* (rs11880261; MAF = 35%; $r^2 = 0.002$, $D' = 0.47$ in the Finnish 1000 Genomes samples) with the common allele associated with lower *AKT2* expression levels (METSIM $P = 6.9 \times 10^{-14}$; EuroBATS $P = 2.3 \times 10^{-8}$; GTEx $P = 0.08$) (Supplementary Figure S15). No association was detected

between rs11880261 and FI levels, suggesting that the common variant eQTL does not drive the initial FI association (Supplementary Note 4; Supplementary Table 10).

Discussion

Meta-analyses of exome sequence and array genotyping data in up to 38,339 normoglycemic individuals enabled the discovery, characterization, and functional validation of a FI association with a low-frequency *AKT2* coding variant. Rare, penetrant variants in genes encoding components of the insulin signaling pathway, including *AKT2*, cause monogenic but heterogeneous glycemic disorders (45). In parallel, common alleles in or near many of these genes impact FI levels—the *AKT2* Pro50Thr association shows an effect 5 to 10 times larger than those of these previous published associations (3). This discovery expands both the known genetic architecture of glucose homeostasis and the allelic spectrum for *AKT2* coding variants associated with glucose homeostasis into the low-frequency range, and highlights the effects of both locus and allelic heterogeneity (Figure 4).

Individuals of Finnish ancestry drove the *AKT2* Pro50Thr association signal. This demonstrates the value of association studies in different ancestries where frequencies of rare alleles may increase due to selective pressure or stochastic changes from population bottlenecks and genetic drift. The allele associated with increased FI most likely rose to a higher frequency due to genetic drift and exists within the spectrum of rare and low-frequency variation observed in Finland, the excess of which facilitates the study of complex trait associations (46).

While the *AKT2* Pro50Thr allele shows a strong effect on all of the insulin measures and modest increased type 2 diabetes risk (OR=1.05) we see no effect on any of the glucose measures in individuals without diabetes. Due to the effects of both type 2 diabetes and its treatment on glucose homeostasis, we have not tested genetic associations of FG and FI in individuals with type 2 diabetes, although we observed a diabetic individual homozygous for P50T with extreme FI and FG levels. The mechanism for such heterogeneous effects is unclear and detailed *in vivo* physiological studies are needed.

We leveraged similar findings to generate hypotheses for future work on *AKT2* and downstream targets to further illuminate tissue-specific mechanisms. All reported carriers of the lipodystrophy causing *AKT2* Arg274His allele are hyperinsulinemic, and three of the four carriers have diabetes mellitus (39). These observations are similar to the ones made for *TBC1D4* (which encodes a protein that acts as a substrate immediately downstream of *AKT2* in the PI3K pathway). In *TBC1D4* a population specific, protein-truncating variant (Arg684Ter) is associated with increased type 2 diabetes risk (OR = 10.3), increased postprandial glucose and insulin levels, and a modest decrease in FI and FG levels (6) (Figure 4). Another stop codon allele in *TBC1D4*, Arg363Ter that is rare (not observed in ExAC) has been reported with a modest elevation in FI levels but extreme postprandial hyperinsulinemia and acanthosis nigricans (47). siRNA-mediated gene knock-down of *AKT2* in human primary myotubes completely abolishes insulin action on glucose uptake and glycogen synthesis (48), which highlights the importance of an intact AKT2-TBC1D4 signaling pathway in the regulation of insulin sensitivity in humans. *TBC1D4* is ubiquitously expressed with adipose and skeletal muscle tissue ranking among the tissues

with highest expression in GTEX. *TBC1D4* Arg363Ter seems to have an effect in adipocytes (47), while Arg684Ter falls in an exon that is exclusively expressed in skeletal and heart muscle (6; 49). This is a likely cause of the *TBC1D4* Arg684Ter tissue specificity, which appears to differ from the other *TBC1D4* Arg363Ter variant as well as the *AKT2* variants.

The phenotypes exhibited by carriers of rare, penetrant *AKT2* alleles reflect differential *AKT2* activation with kinetically inactivating variants resulting in hyperinsulinemia and lipodystrophy while kinetically activating variants lead to hypoglycemia (37–39). The decrease of cellular proliferation we observe demonstrates that the downstream signaling changes caused by *AKT2*-Thr50 are sufficient in hepatocytes to impair *AKT2* function at the cellular level while maintaining varying portions of regulatory capacity. Along with the observed association with increased fasting insulin levels in human populations, these results support *AKT2* Pro50Thr as a *partial* loss-of-function variant. The inactivating *AKT2* Pro50Thr variant contrasts with the known activating *AKT2* Glu17Lys mutation and showcases bidirectional effects within the PH domain of *AKT2*. While the Pro50 residue is conserved in *AKT2* throughout all vertebrates, the variant lies within the PH domain that is not conserved between *AKT* isoforms (Figure 2). These residues, harboring the Pro50 variant, may functionally distinguish *AKT2* from *AKT1* and *AKT3*. Although *AKT* isoforms are activated in the same mechanism within the PI3K pathway downstream of insulin, the *Akt2*^{-/-} mouse is the only knockout of the gene family to be characterized by insulin resistance and diabetes (35; 50–52). A deeper understanding of what makes the *AKT2* isoform distinct could offer potential sites for therapeutic intervention and enable more targeted approaches to disease prevention.

Supplementary Material

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Authors

Alisa Manning^{#1,2,3}, Heather M Highland^{#4,5}, Jessica Gasser^{#1}, Xueling Sim^{#6,7}, Taru Tukiainen^{#1,8,9}, Pierre Fontanillas^{#1,10}, Niels Grarup¹¹, Manuel A Rivas¹², Anubha Mahajan¹², Adam E Locke⁶, Pablo Cingolani^{13,14}, Tune H Pers^{1,11,15,16}, Ana Viñuela^{17,18,19}, Andrew A Brown^{20,21}, Ying Wu²², Jason Flannick^{1,23}, Christian Fuchsberger⁶, Eric R Gamazon^{24,25}, Kyle J Gaulton^{12,26}, Hae Kyung Im²⁴, Tanya M Teslovich⁶, Thomas W Blackwell⁶, Jette Bork-Jensen¹¹, Noël P Burt¹, Yuhui Chen¹², Todd Green¹, Christopher Hartl¹, Hyun Min Kang⁶, Ashish Kumar^{12,27}, Claes Ladenvall²⁸, Clement Ma⁶, Loukas Moutsianas¹², Richard D Pearson¹², John R B Perry^{12,29,30}, N William Rayner^{12,31,32}, Neil R Robertson^{12,31}, Laura J Scott⁶, Martijn van de Bunt^{12,31}, Johan G Eriksson^{33,34,35,36,37}, Antti Jula³⁷, Seppo Koskinen³⁷, Terho Lehtimäki³⁸, Aarno Palotie^{1,2,39}, Olli T Raitakari^{40,41}, Suzanne BR Jacobs¹, Jennifer Wessel^{42,43}, Audrey Y Chu⁴⁴, Robert A Scott³⁰, Mark O Goodarzi^{45,46}, Christine Blancher⁴⁷, Gemma Buck⁴⁷, David Buck⁴⁷, Peter S Chines⁴⁸, Stacey Gabriel¹, Anette P Gjesing¹¹, Christopher J Groves³¹, Mette Hollensted¹¹, Jeroen R Huyghe⁶, Anne U Jackson⁶, Goo Jun⁶, Johanne Marie Justesen¹¹, Massimo Mangino⁴⁹, Jacquelyn Murphy¹, Matt Neville³¹, Robert

Onofrio¹, Kerrin S Small⁴⁹, Heather M Stringham⁶, Joseph Trakalo⁴⁷, Eric Banks¹, Jason Carey¹, Mauricio O Carneiro¹, Mark DePristo¹, Yossi Farjoun¹, Timothy Fennell¹, Jacqueline I Goldstein^{1,8}, George Grant¹, Martin Hrabé de Angelis^{50,51,52}, Jared Maguire¹, Benjamin M Neale^{1,8}, Ryan Poplin¹, Shaun Purcell^{1,2,53}, Thomas Schwarzmayr⁵⁴, Khalid Shakir¹, Joshua D Smith⁵⁵, Tim M Strom^{54,56}, Thomas Wieland⁵⁴, Jaana Lindstrom⁵⁷, Ivan Brandslund^{58,59}, Cramer Christensen⁶⁰, Gabriela L Surdulescu⁴⁹, Timo A Lakka^{61,62,63}, Alex S F Doney⁶⁴, Peter Nilsson⁶⁵, Nicholas J Wareham³⁰, Claudia Langenberg³⁰, Tibor V Varga⁶⁶, Paul W Franks^{66,67,68}, Olov Rolandsson⁶⁸, Anders H Rosengren²⁸, Vidya S Farook⁶⁹, Farook Thameem⁷⁰, Sobha Puppala⁶⁹, Satish Kumar⁶⁹, Donna M Lehman⁷⁰, Christopher P Jenkinson^{70,71}, Joanne E Curran⁶⁹, Daniel Esten Hale⁷², Sharon P Fowler⁷⁰, Rector Arya⁷², Ralph A DeFronzo⁷⁰, Hanna E Abboud⁷⁰, Ann-Christine Syvänen⁷³, Pamela J Hicks^{74,75,76}, Nicholette D Palmer^{74,75,76}, Maggie C Y Ng^{74,75}, Donald W Bowden^{74,75,76}, Barry I Freedman⁷⁷, Tõnu Esko^{1,9,78,79}, Reedik Mägi⁷⁹, Lili Milani⁷⁹, Evelin Mihailov⁷⁹, Andres Metspalu⁷⁹, Narisu Narisu⁴⁸, Leena Kinnunen³⁷, Lori L Bonnycastle⁴⁸, Amy Swift⁴⁸, Dorota Pasko²⁹, Andrew R Wood²⁹, João Fadista²⁸, Toni I Pollin⁸⁰, Nir Barzilai⁸¹, Gil Atzmon⁸¹, Benjamin Glaser⁸², Barbara Thorand^{51,83}, Konstantin Strauch^{84,85}, Annette Peters^{51,83,86}, Michael Roden^{87,88}, Martina Müller-Nurasyid^{84,85,86,89}, Liming Liang^{90,91}, Jennifer Kriebel^{51,83,92}, Thomas Illig^{92,93,94}, Harald Grallert^{51,83,92}, Christian Gieger⁸⁴, Christa Meisinger⁸³, Lars Lannfelt⁹⁵, Solomon K Musani⁹⁶, Michael Griswold⁹⁷, Herman A Taylor Jr⁹⁸, Gregory Wilson Sr⁹⁹, Adolfo Correa⁹⁸, Heikki Oksa¹⁰⁰, William R Scott¹⁰¹, Uzma Afzal¹⁰¹, Sian-Tsung Tan^{102,103}, Marie Loh^{101,104,105}, John C Chambers^{101,103,106}, Jobanpreet Sehmi^{102,103}, Jaspal Singh Kooner¹⁰², Benjamin Lehne¹⁰¹, Yoon Shin Cho¹⁰⁷, Jong-Young Lee¹⁰⁸, Bok-Ghee Han¹⁰⁹, Annemari Käräjämäki^{110,111}, Qibin Qi^{67,112}, Lu Qi^{67,113}, Jinyan Huang⁹⁰, Frank B Hu^{67,90}, Olle Melander¹¹⁴, Marju Orho-Melander¹¹⁵, Jennifer E Below¹¹⁶, David Aguilar¹¹⁷, Tien Yin Wong^{118,119}, Jianjun Liu^{7,120}, Chiea-Chuen Khor^{7,118,119,120,121}, Kee Seng Chia⁷, Wei Yen Lim⁷, Ching-Yu Cheng^{7,118,119,122}, Edmund Chan¹²³, E Shyong Tai^{7,123,124}, Tin Aung^{118,119}, Allan Linneberg^{125,126,127}, Bo Isomaa^{128,129}, Thomas Meitinger^{54,56,86}, Tiinamaija Tuomi^{129,130}, Liisa Hakaste³⁵, Jasmina Kravic²⁸, Marit E Jørgensen¹³¹, Torsten Lauritzen¹³², Panos Deloukas³², Kathleen E Stirrups^{133,134}, Katharine R Owen^{31,135}, Andrew J Farmer¹³⁶, Timothy M Frayling²⁹, Stephen P O'Rahilly¹³⁷, Mark Walker¹³⁸, Jonathan C Levy³¹, Dylan Hodgkiss⁴⁹, Andrew T Hattersley¹³⁹, Teemu Kuulasmaa¹⁴⁰, Alena Stan áková¹⁴⁰, Inês Barroso^{32,137}, Dwaipayan Bharadwaj¹⁴¹, Juliana Chan^{142,143,144}, Giriraj R Chandak¹⁴⁵, Mark J Daly⁸, Peter J Donnelly^{12,146}, Shah B Ebrahim¹⁴⁷, Paul Elliott^{101,148}, Tasha Fingerlin¹⁴⁹, Philippe Froguel¹⁵⁰, Cheng Hu¹⁵¹, Weiping Jia¹⁵¹, Ronald C W Ma^{142,143,144}, Gilean McVean¹², Taesung Park^{152,153}, Dorairaj Prabhakaran¹⁴⁷, Manjinder Sandhu^{32,154}, James Scott¹⁰², Rob Sladek^{14,155,156}, Nikhil Tandon¹⁵⁷, Yik Ying Teo^{7,158,159}, Eleftheria Zeggini³², Richard M Watanabe^{160,161,162}, Heikki A Koistinen^{37,163,164}, Y Antero Kesaniemi¹⁶⁵, Matti Uusitupa¹⁶⁶, Timothy D Spector⁴⁹, Veikko Salomaa³⁷, Rainer Rauramaa¹⁶⁷, Colin N A Palmer¹⁶⁸, Inga Prokopenko^{12,31,169}, Andrew D Morris¹⁷⁰, Richard N Bergman¹⁷¹, Francis S Collins⁴⁸, Lars Lind¹⁷², Erik

Ingelsson^{173,174}, Jaakko Tuomilehto^{57,175,176,177}, Fredrik Karpe^{31,135}, Leif Groop²⁸, Torben Jørgensen^{125,178}, Torben Hansen^{11,179}, Oluf Pedersen¹¹, Johanna Kuusisto^{140,180}, Gonçalo Abecasis⁶, Graeme I Bell¹⁸¹, John Blangero⁶⁹, Nancy J Cox²⁴, Ravindranath Duggirala⁶⁹, Mark Seielstad^{182,183}, James G Wilson¹⁸⁴, Josee Dupuis^{185,186}, Samuli Ripatti^{20,39,187}, Craig L Hanis¹¹⁶, Jose C Florez^{1,2,3,188}, Karen L Mohlke²², James B Meigs^{1,3,189}, Markku Laakso^{140,180}, Andrew P Morris^{12,79,190}, Michael Boehnke⁶, David Altshuler^{1,3,9,23,188,191}, Mark I McCarthy^{12,31,135}, Anna L Gloyn^{12,31,135,&}, and Cecilia M Lindgren^{1,12,192,&}

Affiliations

¹Program in Medical and Population Genetics, Broad Institute, Cambridge, Massachusetts, USA ²Center for Human Genetic Research, Department of Medicine, Massachusetts General Hospital, Boston, Massachusetts, USA ³Department of Medicine, Harvard Medical School, Boston, Massachusetts, USA ⁴Human Genetics Center, The University of Texas Graduate School of Biomedical Sciences at Houston, The University of Texas Health Science Center at Houston, Houston, Texas, USA ⁵Department of Epidemiology, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, USA ⁶Department of Biostatistics and Center for Statistical Genetics, School of Public Health, University of Michigan, Ann Arbor, Michigan, USA ⁷Saw Swee Hock School of Public Health, National University of Singapore, National University Health System, Singapore ⁸Analytic and Translational Genetics Unit, Department of Medicine, Massachusetts General Hospital, Boston, Massachusetts, USA ⁹Department of Genetics, Harvard Medical School, Boston, Massachusetts, USA ¹⁰23andMe, Mountain View, California, USA ¹¹The Novo Nordisk Foundation Center for Basic Metabolic Research, Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark ¹²Wellcome Trust Centre for Human Genetics, Nuffield Department of Medicine, University of Oxford, Oxford, UK ¹³School of Computer Science, McGill University, Montreal, Quebec, Canada ¹⁴McGill University and Génome Québec Innovation Centre, Montreal, Quebec, Canada ¹⁵Divisions of Endocrinology and Genetics and Center for Basic and Translational Obesity Research, Boston Children's Hospital, Boston, Massachusetts, USA ¹⁶Department of Epidemiology Research, Statens Serum Institut, Copenhagen, Denmark ¹⁷Twin Research and Genetic Epidemiology, King's College London, London, UK ¹⁸Department of Genetic Medicine and Development, University of Geneva Medical School, Geneva, Switzerland ¹⁹Institute of Genetics and Genomics in Geneva, University of Geneva, Geneva, Switzerland ²⁰Wellcome Trust Sanger Institute, Hinxton, Cambridge, UK ²¹NORMENT, KG Jebsen Center for Psychosis Research, Division of Mental Health and Addiction, Oslo University Hospital, Oslo, Norway ²²Department of Genetics, University of North Carolina, Chapel Hill, North Carolina, USA ²³Department of Molecular Biology, Massachusetts General Hospital, Boston, Massachusetts, USA ²⁴Department of Medicine, Section of Genetic Medicine, The University of Chicago, Chicago, Illinois, USA ²⁵Academic Medical Center, University of Amsterdam, Amsterdam, The Netherlands ²⁶Department of Pediatrics, University of California San Diego, La Jolla, California, USA ²⁷Chronic Disease Epidemiology, Swiss

Tropical and Public Health Institute, University of Basel, Basel, Switzerland
²⁸Department of Clinical Sciences, Diabetes and Endocrinology, Lund University Diabetes Centre, Malmö, Sweden ²⁹Genetics of Complex Traits, University of Exeter Medical School, University of Exeter, Exeter, UK ³⁰MRC Epidemiology Unit, Institute of Metabolic Science, University of Cambridge, Cambridge, UK ³¹Oxford Centre for Diabetes, Endocrinology and Metabolism, Radcliffe Department of Medicine, University of Oxford, Oxford, UK ³²Department of Human Genetics, Wellcome Trust Sanger Institute, Hinxton, Cambridgeshire, UK ³³Department of General Practice and Primary Healthcare, University of Helsinki, Helsinki, Finland ³⁴Unit of General Practice, Helsinki University Central Hospital, Finland ³⁵Folkhälsan Research Center, Helsinki, Finland ³⁶Vaasa Central Hospital, Vaasa, Finland ³⁷Department of Health, National Institute of Health and Welfare, Helsinki, Finland ³⁸Department of Clinical Chemistry, Fimlab Laboratories, University of Tampere School of Medicine, Tampere, Finland ³⁹Institute for Molecular Medicine Finland, University of Helsinki, Helsinki, Finland ⁴⁰Department of Clinical Physiology and Nuclear Medicine, Turku University Hospital, Turku, Finland ⁴¹Research Centre of Applied and Preventive Cardiovascular Medicine, University of Turku, Turku, Finland ⁴²Department of Epidemiology, Fairbanks School of Public Health, Indianapolis, IN, USA ⁴³Department of Medicine, Indiana University School of Medicine, Indianapolis, IN, USA ⁴⁴Division of Preventive Medicine, Brigham and Women's Hospital, Boston, MA, USA ⁴⁵Division of Endocrinology, Diabetes and Metabolism, Department of Medicine, Cedars-Sinai Medical Center, Los Angeles, California, USA ⁴⁶Department of Biomedical Sciences, Cedars-Sinai Medical Center, Los Angeles, California, USA ⁴⁷High Throughput Genomics, Oxford Genomics Centre, Wellcome Trust Centre for Human Genetics, Nuffield Department of Medicine, University of Oxford, Oxford, UK ⁴⁸National Human Genome Research Institute, National Institutes of Health, Bethesda, Maryland, USA ⁴⁹Department of Twin Research and Genetic Epidemiology, King's College London, London, UK ⁵⁰Institute of Experimental Genetics, Helmholtz Zentrum München, German Research Center for Environmental Health, Neuherberg, Germany ⁵¹German Center for Diabetes Research (DZD), Neuherberg, Germany ⁵²Chair of Experimental Genetics, School of Life Science Weihenstephan, Technische Universität München, Freising, Germany ⁵³Department of Psychiatry, Icahn Institute for Genomics and Multiscale Biology, Icahn School of Medicine at Mount Sinai, New York, USA ⁵⁴Institute of Human Genetics, Helmholtz Zentrum München, German Research Center for Environmental Health, Neuherberg, Germany ⁵⁵Department of Genome Sciences, University of Washington School of Medicine, Seattle, Washington, USA ⁵⁶Institute of Human Genetics, Technische Universität München, Munich, Germany ⁵⁷Diabetes Prevention Unit, National Institute for Health and Welfare, Helsinki, Finland ⁵⁸Institute of Regional Health Research, University of Southern Denmark, Odense, Denmark ⁵⁹Department of Clinical Biochemistry, Vejle Hospital, Vejle, Denmark ⁶⁰Department of Internal Medicine and Endocrinology, Vejle Hospital, Vejle, Denmark ⁶¹Institute of Biomedicine, Physiology, University of Eastern Finland, Kuopio, Finland ⁶²Kuopio Research Institute of Exercise Medicine, Kuopio, Finland

⁶³Department of Clinical Physiology and Nuclear Medicine, Kuopio University Hospital, Kuopio, Finland ⁶⁴Division of Cardiovascular and Diabetes Medicine, Medical Research Institute, Ninewells Hospital and Medical School, Dundee, UK ⁶⁵Department of Clinical Sciences, Medicine, Lund University, Malmö, Sweden ⁶⁶Department of Clinical Sciences, Lund University Diabetes Centre, Genetic and Molecular Epidemiology Unit, Lund University, Malmö, Sweden ⁶⁷Department of Nutrition, Harvard School of Public Health, Boston, Massachusetts, USA ⁶⁸Department of Public Health and Clinical Medicine, Umeå University, Umeå, Sweden ⁶⁹Department of Genetics, Texas Biomedical Research Institute, San Antonio, Texas, USA ⁷⁰Department of Medicine, University of Texas Health Science Center, San Antonio, Texas, USA ⁷¹Research, South Texas Veterans Health Care System, San Antonio, Texas, USA ⁷²Department of Pediatrics, University of Texas Health Science Center, San Antonio, Texas, USA ⁷³Department of Medical Sciences, Molecular Medicine and Science for Life Laboratory, Uppsala University, Uppsala, Sweden ⁷⁴Center for Genomics and Personalized Medicine Research, Wake Forest School of Medicine, Winston-Salem, North Carolina, USA ⁷⁵Center for Diabetes Research, Wake Forest School of Medicine, Winston-Salem, North Carolina, USA ⁷⁶Department of Biochemistry, Wake Forest School of Medicine, Winston-Salem, North Carolina, USA ⁷⁷Department of Internal Medicine, Section on Nephrology, Wake Forest School of Medicine, Winston-Salem, North Carolina, USA ⁷⁸Division of Endocrinology, Boston Children's Hospital, Boston, Massachusetts, USA ⁷⁹Estonian Genome Center, University of Tartu, Tartu, Estonia ⁸⁰Department of Medicine, Program in Personalized and Genomic Medicine, University of Maryland, Baltimore, Maryland, USA ⁸¹Departments of Medicine and Genetics, Albert Einstein College of Medicine, New York, USA ⁸²Endocrinology and Metabolism Service, Hadassah-Hebrew University Medical Center, Jerusalem, Israel ⁸³Institute of Epidemiology II, Helmholtz Zentrum München, German Research Center for Environmental Health, Neuherberg, Germany ⁸⁴Institute of Genetic Epidemiology, Helmholtz Zentrum München, German Research Center for Environmental Health, Neuherberg, Germany ⁸⁵Institute of Medical Informatics, Biometry and Epidemiology, Chair of Genetic Epidemiology, Ludwig-Maximilians-Universität, Munich, Germany ⁸⁶Deutsches Forschungszentrum für Herz-Kreislaferkrankungen (DZHK), Partner Site Munich Heart Alliance, Munich, Germany ⁸⁷Institute of Clinical Diabetology, German Diabetes Center, Leibniz Center for Diabetes Research at Heinrich Heine University, Düsseldorf, Germany ⁸⁸German Center for Diabetes Research, Partner Düsseldorf, Germany ⁸⁹Department of Medicine I, University Hospital Grosshadern, Ludwig-Maximilians-Universität, Munich, Germany ⁹⁰Department of Epidemiology, Harvard School of Public Health, Boston, Massachusetts, USA ⁹¹Department of Biostatistics, Harvard School of Public Health, Boston, Massachusetts, USA ⁹²Research Unit of Molecular Epidemiology, Helmholtz Zentrum München, German Research Center for Environmental Health, Neuherberg, Germany ⁹³Hannover Unified Biobank, Hannover Medical School, Hannover, Germany ⁹⁴Institute of Human Genetics, Hannover Medical School, Hannover, Germany ⁹⁵Department of Public Health and Caring Sciences, Geriatrics,

Uppsala University, Uppsala, Sweden ⁹⁶Jackson Heart Study, University of Mississippi Medical Center, Jackson, Mississippi, USA ⁹⁷Center of Biostatistics and Bioinformatics, University of Mississippi Medical Center, Jackson, Mississippi, USA ⁹⁸Department of Medicine, University of Mississippi Medical Center, Jackson, Mississippi, USA ⁹⁹College of Public Services, Jackson State University, Jackson, Mississippi, USA ¹⁰⁰Pirkanmaa Hospital District, Tampere, Finland ¹⁰¹Department of Epidemiology and Biostatistics, Imperial College London, London, UK ¹⁰²National Heart and Lung Institute, Cardiovascular Sciences, Hammersmith Campus, Imperial College London, London, UK ¹⁰³Department of Cardiology, Ealing Hospital NHS Trust, Southall, Middlesex, UK ¹⁰⁴Institute of Health Sciences, University of Oulu, Oulu, Finland ¹⁰⁵Translational Laboratory in Genetic Medicine (TLGM), Agency for Science, Technology and Research (A*STAR), Singapore ¹⁰⁶Imperial College Healthcare NHS Trust, Imperial College London, London, UK ¹⁰⁷Department of Biomedical Science, Hallym University, Chuncheon, Republic of Korea ¹⁰⁸Ministry of Health and Welfare, Seoul, Republic of Korea ¹⁰⁹Center for Genome Science, Korea National Institute of Health, Chungcheongbuk-do, Republic of Korea ¹¹⁰Vasa Health Care Center, Vaasa, Finland ¹¹¹Department of Primary Health Care, Vasa Central Hospital, Vasa, Finland ¹¹²Department of Epidemiology and Population Health, Albert Einstein College of Medicine, New York, USA ¹¹³Channing Division of Network Medicine, Department of Medicine, Brigham and Women's Hospital and Harvard Medical School, Boston, Massachusetts, USA ¹¹⁴Department of Clinical Sciences, Hypertension and Cardiovascular Disease, Lund University, Malmö, Sweden ¹¹⁵Department of Clinical Sciences, Diabetes and Cardiovascular Disease, Genetic Epidemiology, Lund University, Malmö, Sweden ¹¹⁶Human Genetics Center, School of Public Health, The University of Texas Health Science Center at Houston, Houston, Texas, USA ¹¹⁷Cardiovascular Division, Baylor College of Medicine, Houston, Texas, USA ¹¹⁸Singapore Eye Research Institute, Singapore National Eye Centre, Singapore ¹¹⁹Department of Ophthalmology, Yong Loo Lin School of Medicine, National University of Singapore, National University Health System, Singapore ¹²⁰Division of Human Genetics, Genome Institute of Singapore, A*STAR, Singapore ¹²¹Department of Paediatrics, Yong Loo Lin School of Medicine, National University of Singapore, National University Health System, Singapore ¹²²Centre for Quantitative Medicine, Office of Clinical Sciences, Duke-NUS Graduate Medical School Singapore, Singapore ¹²³Department of Medicine, Yong Loo Lin School of Medicine, National University of Singapore, National University Health System, Singapore ¹²⁴Cardiovascular & Metabolic Disorders Program, Duke-NUS Graduate Medical School Singapore, Singapore ¹²⁵Research Centre for Prevention and Health, Glostrup University Hospital, Glostrup, Denmark ¹²⁶Department of Clinical Experimental Research, Rigshospitalet, Glostrup, Denmark ¹²⁷Department of Clinical Medicine, Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark ¹²⁸Department of Social Services and Health Care, Jakobstad, Finland ¹²⁹Folkhälsan Research Centre, Helsinki, Finland ¹³⁰Department of Endocrinology, Helsinki University Central Hospital, Helsinki, Finland ¹³¹Steno Diabetes Center, Gentofte, Denmark ¹³²Department of Public

Health, Section of General Practice, Aarhus University, Aarhus, Denmark ¹³³William Harvey Research Institute, Barts and The London School of Medicine and Dentistry, Queen Mary University of London, London, UK ¹³⁴Department of Haematology, University of Cambridge, Cambridge, UK ¹³⁵Oxford NIHR Biomedical Research Centre, Oxford University Hospitals Trust, Oxford, UK ¹³⁶Department of Primary Care Health Sciences, University of Oxford, Oxford, UK ¹³⁷Metabolic Research Laboratories, Institute of Metabolic Science, University of Cambridge, Cambridge, UK ¹³⁸The Medical School, Institute of Cellular Medicine, University of Newcastle, Newcastle, UK ¹³⁹University of Exeter Medical School, University of Exeter, Exeter, UK ¹⁴⁰Faculty of Health Sciences, Institute of Clinical Medicine, Internal Medicine, University of Eastern Finland, Kuopio, Finland ¹⁴¹Functional Genomics Unit, CSIR-Institute of Genomics & Integrative Biology (CSIR-IGIB), New Delhi, India ¹⁴²Department of Medicine and Therapeutics, The Chinese University of Hong Kong, Hong Kong, China ¹⁴³Li Ka Shing Institute of Health Sciences, The Chinese University of Hong Kong, Hong Kong, China ¹⁴⁴Hong Kong Institute of Diabetes and Obesity, The Chinese University of Hong Kong, Hong Kong, China ¹⁴⁵CSIR-Centre for Cellular and Molecular Biology, Hyderabad, Andhra Pradesh, India ¹⁴⁶Department of Statistics, University of Oxford, Oxford, UK ¹⁴⁷Centre for Chronic Disease Control, New Delhi, India ¹⁴⁸MRC-PHE Centre for Environment and Health, Imperial College London, London, UK ¹⁴⁹Department of Epidemiology, Colorado School of Public Health, University of Colorado, Aurora, Colorado, USA ¹⁵⁰Genomics and Molecular Physiology, CNRS (Institut de Biologie de Lille), Lille, France ¹⁵¹Department of Endocrinology and Metabolism, Shanghai Diabetes Institute, Shanghai Jiao Tong University Affiliated Sixth People's Hospital, Shanghai, China ¹⁵²Interdisciplinary Program in Bioinformatics, Seoul National University, Seoul, Republic of Korea ¹⁵³Department of Statistics, Seoul National University, Seoul, Republic of Korea ¹⁵⁴Department of Public Health and Primary Care, Institute of Public Health, University of Cambridge, Cambridge, UK ¹⁵⁵Department of Human Genetics, McGill University, Montreal, Quebec, Canada ¹⁵⁶Division of Endocrinology and Metabolism, Department of Medicine, McGill University, Montreal, Quebec, Canada ¹⁵⁷Department of Endocrinology and Metabolism, All India Institute of Medical Sciences, New Delhi, India ¹⁵⁸Life Sciences Institute, National University of Singapore, Singapore ¹⁵⁹Department of Statistics and Applied Probability, National University of Singapore, Singapore ¹⁶⁰Department of Preventive Medicine, Keck School of Medicine, University of Southern California, Los Angeles, California, USA ¹⁶¹Department of Physiology & Biophysics, Keck School of Medicine, University of Southern California, Los Angeles, California, USA ¹⁶²Diabetes and Obesity Research Institute, Keck School of Medicine, University of Southern California, Los Angeles, California, USA ¹⁶³University of Helsinki and Helsinki University Central Hospital, Department of Medicine and Abdominal Center, Endocrinology, Helsinki, Finland ¹⁶⁴Minerva Foundation Institute for Medical Research, Helsinki, Finland ¹⁶⁵Institute of Clinical Medicine, Department of Medicine, University of Oulu and Medical Research Center, Oulu University Hospital, Oulu, Finland ¹⁶⁶Institute of Public Health and Clinical Nutrition, University of Eastern Finland, Kuopio, Finland

¹⁶⁷Foundation for Research in Health, Exercise and Nutrition, Kuopio Research Institute of Exercise Medicine, Kuopio, Finland ¹⁶⁸Pat Macpherson Centre for Pharmacogenetics and Pharmacogenomics, Medical Research Institute, Ninewells Hospital and Medical School, Dundee, UK ¹⁶⁹Department of Genomics of Common Disease, School of Public Health, Imperial College London, London, UK ¹⁷⁰Clinical Research Centre, Centre for Molecular Medicine, Ninewells Hospital and Medical School, Dundee, UK ¹⁷¹Cedars-Sinai Diabetes and Obesity Research Institute, Los Angeles, California, USA ¹⁷²Department of Medical Sciences, Uppsala University, Uppsala, Sweden ¹⁷³Department of Medical Sciences, Molecular Epidemiology and Science for Life Laboratory, Uppsala University, Uppsala, Sweden ¹⁷⁴Department of Medicine, Division of Cardiovascular Medicine, Stanford University School of Medicine, Stanford, California, USA ¹⁷⁵Center for Vascular Prevention, Danube University Krems, Krems, Austria ¹⁷⁶Diabetes Research Group, King Abdulaziz University, Jeddah, Saudi Arabia ¹⁷⁷Dasman Diabetes Institute, Dasman, 15642 Kuwait ¹⁷⁸Faculty of Medicine, University of Aalborg, Aalborg, Denmark ¹⁷⁹Faculty of Health Sciences, University of Southern Denmark, Odense, Denmark ¹⁸⁰Kuopio University Hospital, Kuopio, Finland ¹⁸¹Departments of Medicine and Human Genetics, The University of Chicago, Chicago, Illinois, USA ¹⁸²Department of Laboratory Medicine & Institute for Human Genetics, University of California, San Francisco, San Francisco, California, USA ¹⁸³Blood Systems Research Institute, San Francisco, California, USA ¹⁸⁴Department of Physiology and Biophysics, University of Mississippi Medical Center, Jackson, Mississippi, USA ¹⁸⁵Department of Biostatistics, Boston University School of Public Health, Boston, Massachusetts, USA ¹⁸⁶National Heart, Lung, and Blood Institute's Framingham Heart Study, Framingham, Massachusetts, USA ¹⁸⁷Hjelt Institute, University of Helsinki, Helsinki, Finland ¹⁸⁸Diabetes Research Center (Diabetes Unit), Department of Medicine, Massachusetts General Hospital, Boston, Massachusetts, USA ¹⁸⁹Division of General Internal Medicine, Massachusetts General Hospital, Boston, Massachusetts, USA ¹⁹⁰Department of Biostatistics, University of Liverpool, Liverpool, UK ¹⁹¹Department of Biology, Massachusetts Institute of Technology, Cambridge, Massachusetts, USA ¹⁹²Li Ka Shing Centre for Health Information and Discovery, The Big Data Institute, University of Oxford, Oxford, UK

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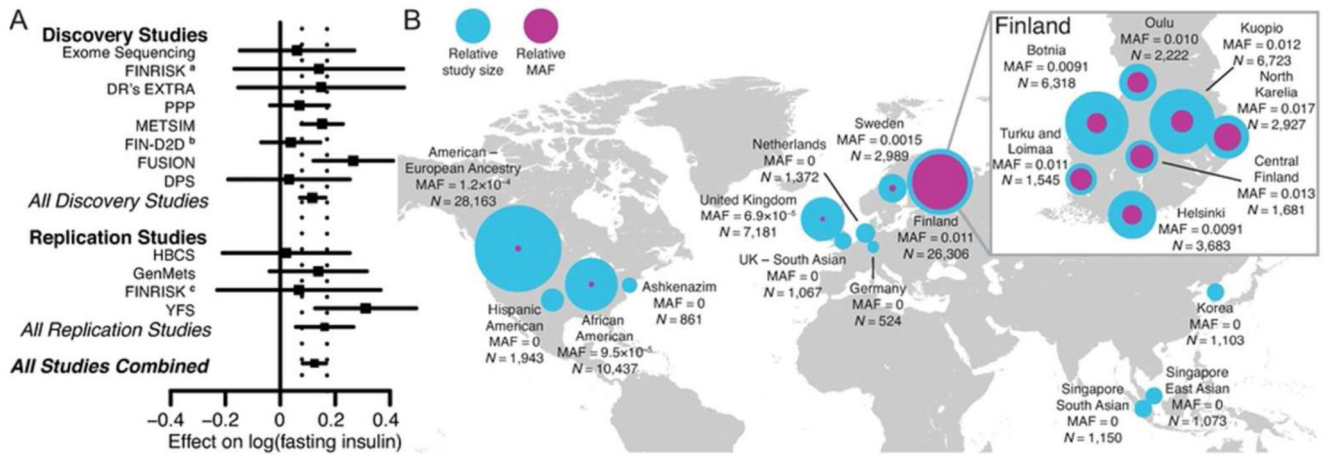


Figure 1. *AKT2* Pro50Thr association with fasting insulin levels.

(a) For each study, the square represents the estimate of the additive genetic effect for the association of the *AKT2* Pro50Thr allele with log-transformed fasting insulin (FI) levels and the horizontal line gives the corresponding 95% confidence interval of the estimate. Inverse-variance meta-analyses were performed for *All Discovery Studies*, *All Replication Studies*, and *All Studies Combined*. The vertical dashed lines indicate the 95% confidence interval for the estimate obtained in the meta-analysis of *All Studies Combined*. (b) Minor allele frequency for each available region and ancestry. Across countries the world, the MAF ranges from 0% to 1.1%. The relative sample sizes (N) for each region/ancestry are displayed with the blue circles and the relative minor allele frequencies of *AKT2* Pro50Thr are displayed with the purple circles, with the size of the circles showing comparative differences. Within Finland (inset), where the MAF ranges from 0.9% to 1.7%, birthplace and study center data were used to show the allele distribution across the country.^a FINRISK 2007; ^b FIN-D2D 2007; ^c FINRISK 1997 and 2002

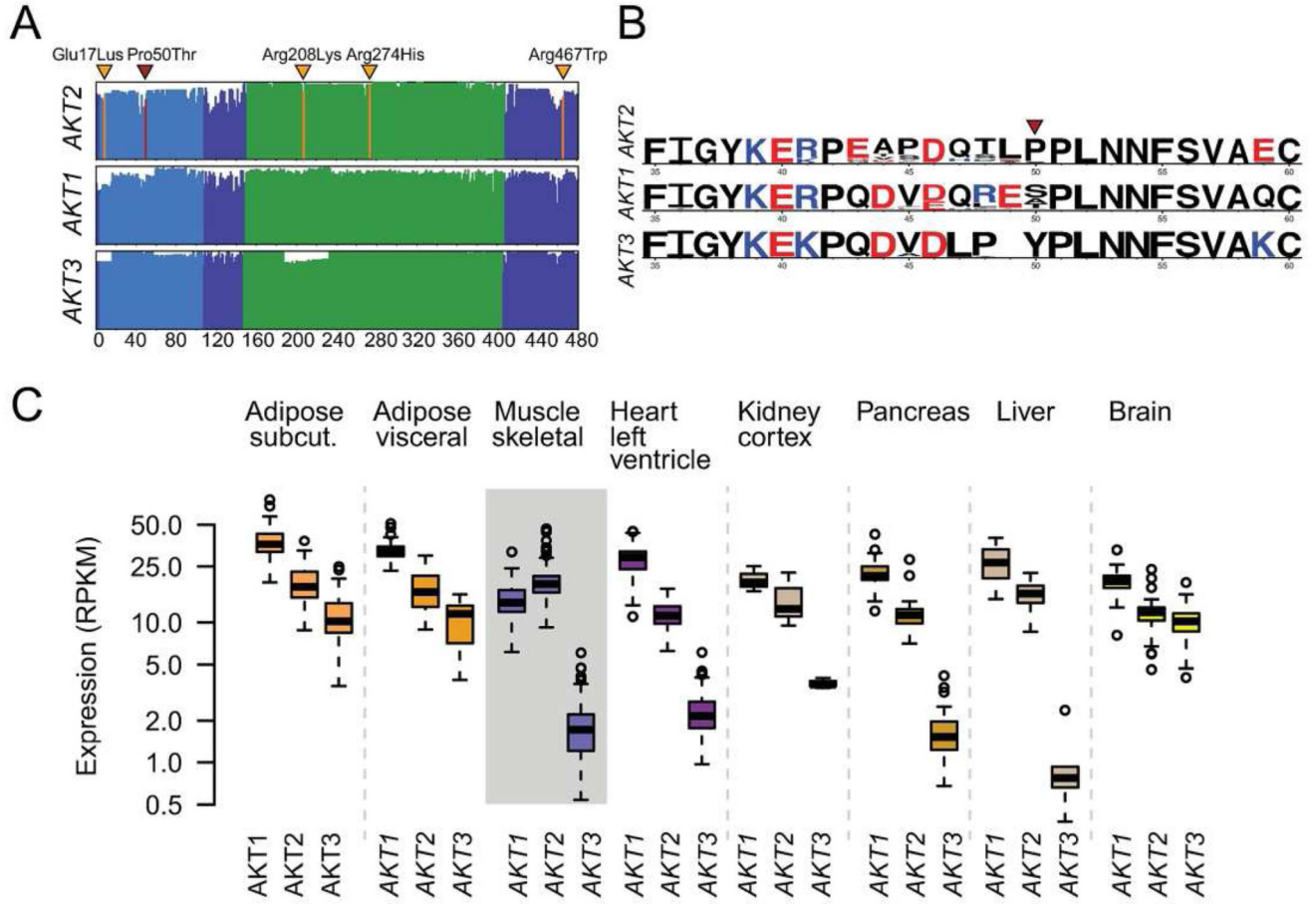


Figure 2. Expression and conservation properties.

(a) Amino acid alignment and conservation of the three AKT proteins in vertebrates. The x axis gives the amino acid position and the height of the lines shows the conservation score across 100 vertebrate genome alignments. The functional domains are the pleckstrin homology (PH) domain (blue) and the kinase domain (green). The position of AKT2 Pro50Thr is shown in red while the locations of the other *AKT2* disease-causing mutations (37–40) are shown in orange: Glu17Lys, Arg208Lys, Arg274His, and Arg467Trp. (b) WebLogo plots of amino acids 35–60 are shown for AKT2, AKT1, and AKT3 contrasting the homology of the three isoforms. The height of letters gives the relative frequency of different amino acids across the 100 vertebrate species, with the colors showing amino acids with similar charge. (c) Expression of *AKT1*, *AKT2*, and *AKT3* in eight insulin-sensitive tissues using RNA sequencing data from the GTEx consortium.

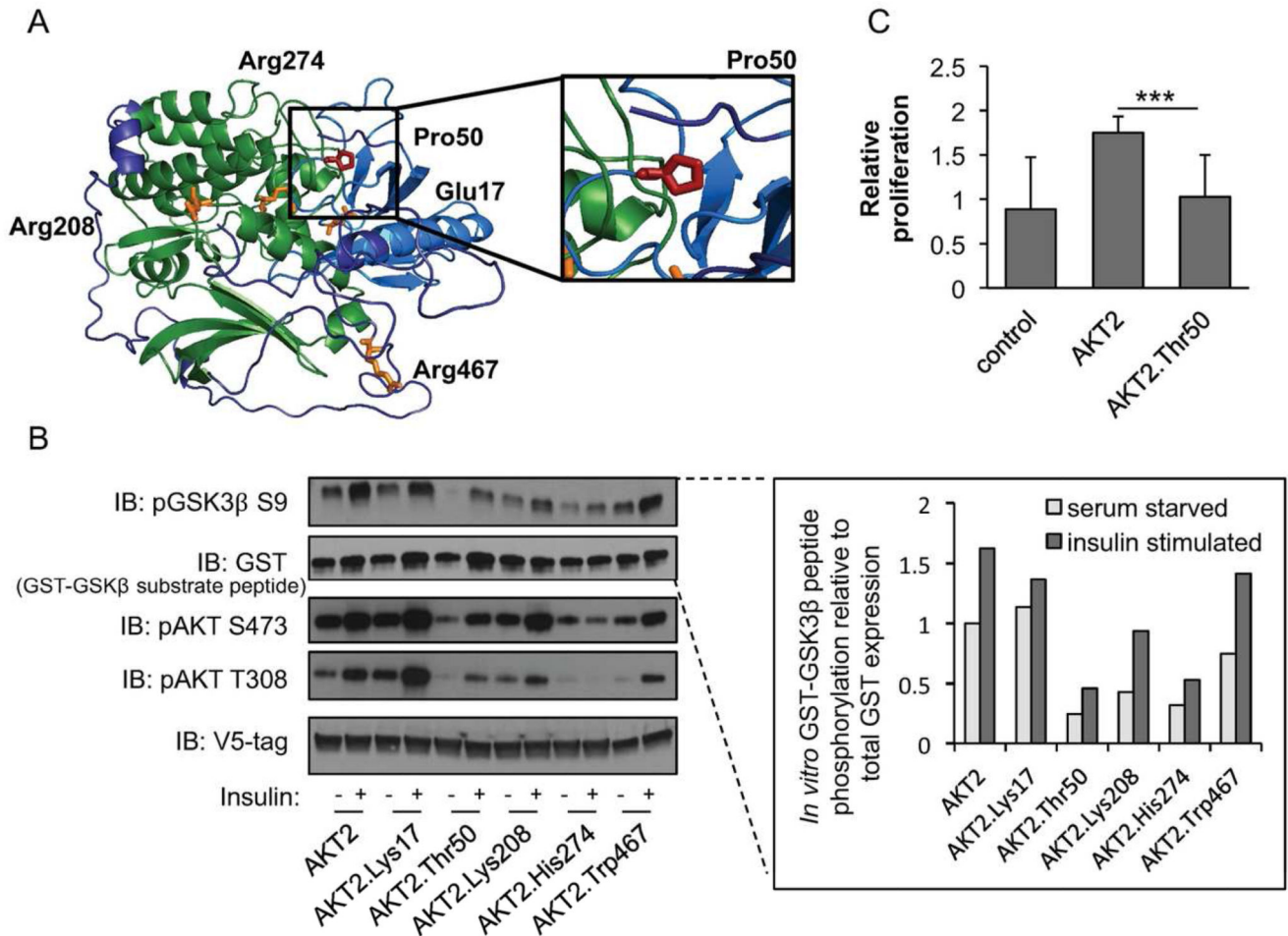


Figure 3. Functional properties of AKT2-Thr50

(a) Predicted protein structure of AKT2. Domain and variants are highlighted as in Figure 2. The relative spatial positioning of the AKT2-Pro50 residue is magnified within the inset. (b) HeLa cells were infected with lentiviral V5-AKT2, V5-AKT2-Lys17, V5-AKT2-Thr50, V5-AKT2-Lys208, V5-AKT2-His274, V5-AKT2-Trp467, starved for 18 hours (white bar), and stimulated for 20 minutes with 100nm insulin (grey bar). V5-tagged AKT2 was isolated from cell lysates with anti-V5 agarose beads and incubated with GSK3β-GST peptide in an *in vitro* kinase (IVK) assay. Quantification of phosphorylated substrate peptide (pGSK3β) relative to total peptide (GST-GSK3β) is shown at the inset. Immunoblots and quantification shown are representative of three independent replicates. Linear model (LM) statistical analyses across all three independent replicates are available in Supplementary Figure S9. The IVK was immunoblotted (IB) with the indicated antibodies. (c) HuH7 cells were infected with lentiviral V5-AKT2, V5-AKT2-Thr50, or control pLX304. At 72 hours relative cellular proliferation was determined with WST-1 assay of HuH7 cells. Error bars represent the standard deviation (SD). *** $P=4.5 \times 10^{-5}$.

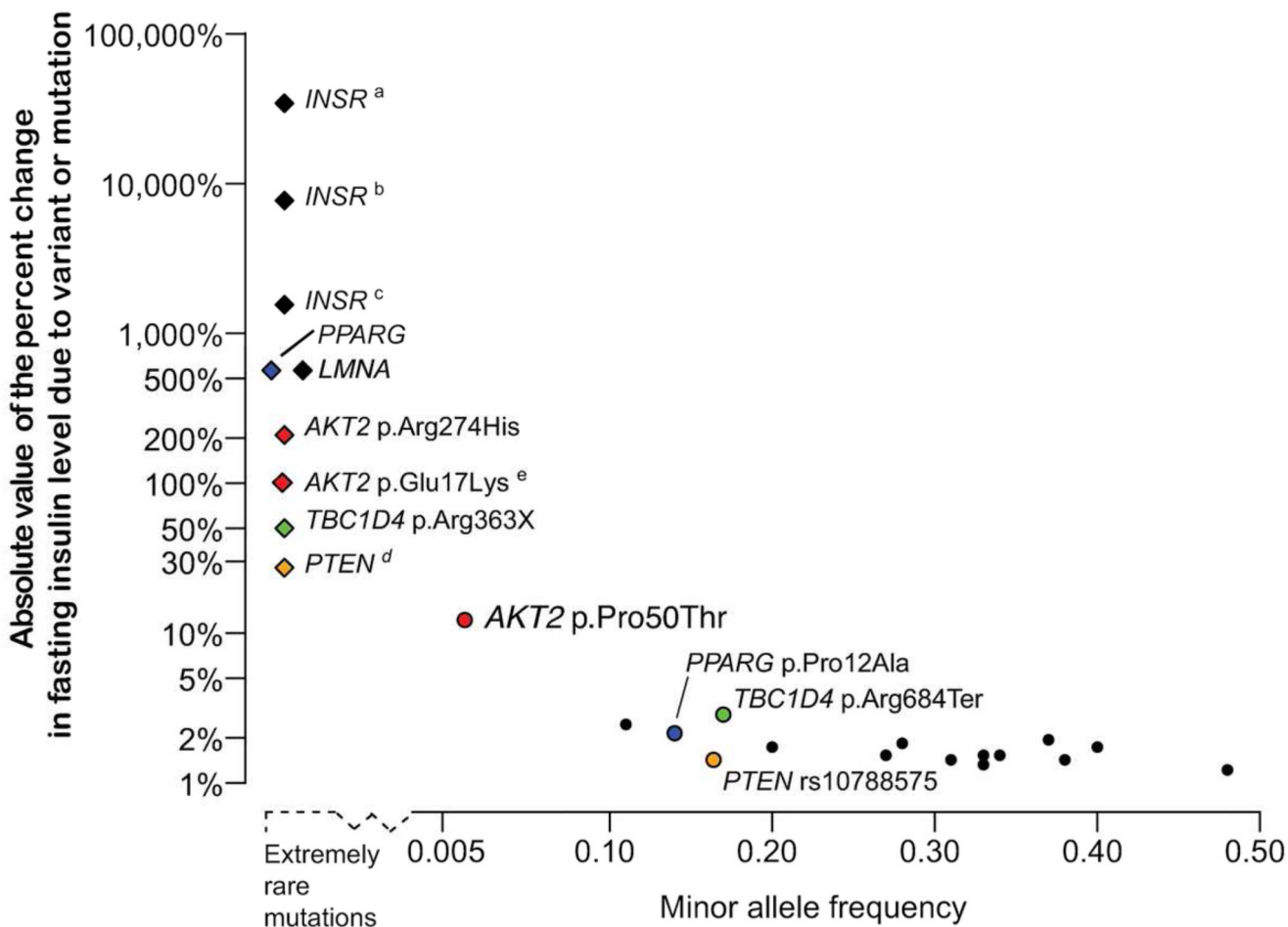


Figure 4. Genetic architecture of rare, low frequency, and common variants associated with FI levels.

In this plot, the absolute values of the percent change in fasting insulin level due to rare monogenic mutations (diamonds) and common genetic variants (circles) are plotted against the minor allele frequency of the variant. The extremely rare monogenic mutations (above the dashed line to the left of the x axis) were observed in 2 to 18 individuals (3; 37–40; 47; 53; 54) with the height of the point indicating the percent change in fasting insulin levels of mutation carriers from 40 pmol/L, an estimate of population mean fasting insulin level. Mutations in *INSR* and *AKT2* p.Arg274His cause compensatory hyperinsulinemia, individuals with *TBC1D4* p.Arg363Ter show normal fasting insulin levels but postprandial hyperinsulinemia, and mutations in *PTEN* cause enhanced insulin sensitivity providing protection against type 2 diabetes. For common variants, the percent change in fasting insulin levels per insulin-increasing allele is plotted above the solid horizontal axis. These observations are from sequencing (6) and array-based GWAS (3). For several genes, the effects from rare mutations can be compared to the effects of common variants in or near the gene: *PPARG* (blue), *TBC1D4* (green), *PTEN* (orange), and *AKT2* (red). ^a Donohue syndrome: Biallelic loss-of-function mutations in *INSR* (54). ^b Rabson-Mendenhall syndrome: Biallelic loss-of-function mutations in *INSR* (54). ^c Post-pubertal severe IR:

Heterozygous or homozygous loss-of-function mutations in *INSR* (54).^d Loss of function *PTEN* mutations cause Cowden Syndrome in which carriers exhibit a *lowered* fasting insulin level (mean=29 pmol/l) compared to matched controls (3).^e Carriers with the *AKT2* p.Glu17Lys mutation were described with hypoinsulinemic hypoketotic hypoglycemia and hemihypertrophy with undetectable serum insulin (37; 38).