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1 Association between LDL-cholesterol lowering genetic variants and risk of type 2 2 diabetes

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81

82 **Abstract**

83

84

85 **Importance:** Low-density lipoprotein (LDL) cholesterol-lowering alleles in or near *NPC1L1*
86 or *HMGCR*, encoding the respective molecular targets of ezetimibe and statins, have
87 previously been used as proxies to study the efficacy of these lipid-lowering drugs. Alleles
88 near *HMGCR* are associated with a higher risk of type 2 diabetes, mimicking the increased
89 incidence of new-onset diabetes associated with statin treatment in randomized clinical trials.
90 It is unknown whether alleles near *NPC1L1* are also associated with the risk of type 2
91 diabetes.

92

93 **Objective:** To investigate whether LDL-lowering alleles in or near *NPC1L1* and other genes
94 encoding current or prospective molecular targets of lipid-lowering therapy (i.e. *HMGCR*,
95 *PCSK9*, *ABCG5/G8*, *LDLR*) are associated with the risk of type 2 diabetes.

96

97 **Design, Setting and Participants:** The associations with type 2 diabetes and coronary artery
98 disease of LDL-lowering genetic variants were investigated in meta-analyses of genetic
99 association studies. Meta-analyses included 50,775 individuals with type 2 diabetes and
100 270,269 controls including three studies and 60,801 individuals with coronary artery disease
101 and 123,504 controls from a published meta-analysis. Data collection took place in Europe
102 and the United States between 1991 and 2016.

103

104 **Exposure:** LDL-lowering alleles in or near *NPC1L1*, *HMGCR*, *PCSK9*, *ABCG5/G8*, *LDLR*.

105

106 **Main Outcomes and Measures:** Odds ratio of type 2 diabetes and coronary artery disease.

107

108 **Results:** LDL-lowering genetic variants at *NPC1L1* were inversely associated with coronary
109 artery disease (odds ratio for a genetically-predicted reduction of 1 mmol/L in LDL
110 cholesterol, 0.61; 95% confidence interval, 0.42-0.88; p=0.008) and directly associated with
111 type 2 diabetes (2.42, 1.70-3.43; p<0.001). The odds ratio of type 2 diabetes for *PCSK9*
112 genetic variants was 1.19 (95% confidence interval, 1.02-1.38, p=0.03). For a given reduction
113 in LDL cholesterol, genetic variants were associated with a similar reduction in coronary
114 artery disease risk (I-squared for heterogeneity in genetic associations=0.0%; p=0.93).
115 However, associations with type 2 diabetes were heterogeneous (I-squared=77.2%; p=0.002),
116 indicating gene-specific associations with metabolic risk for LDL-lowering alleles.

117

118 **Conclusions and Relevance:** In this meta-analysis, exposure to LDL-cholesterol lowering
119 genetic variants in or near *NPC1L1* and other genes was associated with a higher risk of type
120 2 diabetes. These data provide insights into potential adverse effects of LDL cholesterol-
121 lowering therapy.

122 **Key Points**

123

124 • **Question:** Are LDL-cholesterol lowering alleles at *NPC1L1* or other genes associated
125 with the risk of type 2 diabetes?

126

127 • **Findings:** In a meta-analysis of genetic association studies including 50,775
128 individuals with type 2 diabetes and 270,269 controls, LDL-lowering polymorphisms at
129 *NPC1L1* were associated with a statistically significant odds ratio for type 2 diabetes of 2.42
130 per genetically-predicted reduction of 1 mmol/L in LDL cholesterol. LDL-lowering
131 polymorphisms at *HMGCR* and *PCSK9* were also associated with a higher risk of diabetes.

132

133 • **Meaning:** These data provide insights into potential adverse effects of LDL
134 cholesterol-lowering therapy.

135 **Introduction**

136 Treatment with statins, the pharmacological agents of choice for low-density
137 lipoprotein (LDL) cholesterol-lowering therapy in cardiovascular prevention,^{1,2} is associated
138 with weight gain and a higher incidence of new-onset type 2 diabetes.³⁻⁵ Ezetimibe, an
139 inhibitor of the LDL cholesterol transporter Niemann-Pick C1-like 1 (NPC1L1),^{6,7} has been
140 approved as a lipid-lowering agent, but it is unclear whether its use will also be associated
141 with an adverse metabolic risk profile.

142 There is considerable interest in predicting the efficacy and safety of therapeutic targets
143 early in the drug development process. Drug targets with supporting human genetic evidence
144 have been shown to have lower attrition rates during drug development,⁸ while variation in
145 genes encoding drug targets has been used to predict both the efficacy and safety of
146 pharmacological perturbation of those targets.^{9,10} In particular, LDL-lowering alleles in
147 *HMGCR*^{5,11} encoding the molecular target of statins, have been successfully used as genetic
148 proxies to study the effects of these drugs.^{5,11} Furthermore, LDL-lowering alleles at *HMGCR*
149 are associated with higher risk of type 2 diabetes and higher body mass index in genetic
150 studies,⁵ mimicking the safety profile of statins in meta-analyses of randomized clinical
151 trials.³⁻⁵

152 The efficacy of adding ezetimibe to simvastatin in secondary cardiovascular prevention
153 was supported by the IMPROVE-IT trial.^{6,7} Immediately before and after the publication of
154 the trial results, studies were reported describing the use of genetic variants at *NPC1L1* to
155 predict the efficacy of NPC1L1 inhibition in the prevention of coronary events.^{11,12} The
156 purpose of this study was to use naturally-occurring LDL-lowering alleles at *NPC1L1* to
157 investigate the potential associations between NPC1L1 inhibition and the risk of type 2
158 diabetes. LDL-lowering alleles in or near genes encoding other current or prospective
159 molecular targets of LDL-cholesterol lowering therapy were also studied.

160 **Methods**

161 *Study design*

162 The association of LDL-cholesterol lowering polymorphisms near *NPC1L1* with the
163 risk of type 2 diabetes was investigated in meta-analyses of genetic association studies. In
164 addition to *NPC1L1* polymorphisms, the association of LDL-lowering alleles in or near genes
165 encoding other current or prospective molecular targets of LDL-cholesterol lowering
166 therapy¹¹ (i.e. *HMGCR*, *PCSK9*, *ABCG5/G8*, *LDLR*) with type 2 diabetes, coronary artery
167 disease and continuous cardiometabolic traits was also studied. A summary of the studies
168 participating in each analysis is presented in **eTable 1**.

169

170 *Participants*

171 The association of LDL-cholesterol lowering alleles with type 2 diabetes was estimated
172 in a meta-analysis of 50,775 individuals with type 2 diabetes and 270,269 controls from the
173 EPIC-InterAct study¹³, the UK Biobank study¹⁴ and the DIABetes Genetics Replication And
174 Meta-analysis (DIAGRAM).¹⁵ An additional eleven studies (4,496 cases and 50,677 controls)
175 previously reported by Swerdlow and colleagues⁵ were included in analyses of the
176 association with type 2 diabetes of rs12916 in *HMGCR* (**eFigure 1**). The combined
177 association of *NPC1L1* genetic variants in subgroups of age, sex, and body mass index was
178 analyzed in 14,657 unrelated cases of type 2 diabetes and 118,854 controls from EPIC-
179 InterAct and UK Biobank with available individual-level genotyping data.

180 EPIC-InterAct is a case-cohort study nested within the European Prospective
181 Investigation into Cancer and Nutrition (EPIC) study, a cohort study of 500,000 European
182 participants followed-up for an average of 8 years.¹³ Eight of the ten constituent EPIC cohorts
183 agreed to take part in EPIC-InterAct leaving 455,680 participants for screening. Individuals
184 were excluded from EPIC-InterAct if they did not have stored blood (n=109,625) or

185 information on diabetes status (n=5,821; 1.3% of participants screened for inclusion). From
186 the remaining 340,234 participants, 12,403 individuals who developed type 2 diabetes during
187 follow-up constituted the incident case group of EPIC-InterAct and a random group of 16,154
188 individuals free of diabetes at baseline constituted the subcohort group of EPIC-InterAct.¹³
189 Subcohort participants were previously shown to be representative of eligible EPIC
190 participants within each country.¹³ Data on a total of 20,831 participants with available
191 genotyping (with no overlap with DIAGRAM¹⁵) were included in the main analysis, while
192 data on all the 22,494 participants with available genotyping were included in subgroup
193 analyses. Type 2 diabetes status was available in all participants. Individuals without
194 genotype data were excluded from the study. Data collection took place between 1991 and
195 2016. Participant characteristics and genotyping methods have been previously reported in
196 detail¹³ and are summarized in **Table 1 and eTable 2**.

197 UK Biobank is a population-based cohort of 500,000 people aged between 40-69 years
198 who were recruited in 2006-2010 from several centers across the United Kingdom.¹⁴ The
199 association of genetic variants with prevalent type 2 diabetes was estimated in 6,627 cases
200 and 143,765 controls of the UK Biobank dataset who had available genotype data.
201 Genotyping was attempted in 152,770 individuals and failed in only 480 instances (0.3%).
202 Among a total of 152,290 participants with available genotype data, type 2 diabetes status
203 was adjudicated in 150,392 (98.8%) participants. Type 2 diabetes was defined on the basis of
204 self-reported physician diagnosis at nurse interview or digital questionnaire, age at diagnosis
205 > 36 years, and use of oral anti-diabetic medications. Data collection took place between
206 2006 and 2016. Participant characteristics and genotyping information are reported in **Table**
207 **1 and eTable 2**.

208 DIAGRAM is a research consortium that published the largest meta-analysis of
209 genome-wide association studies for type 2 diabetes in individuals of European descent.¹⁵

210 Type 2 diabetes association results were made publicly available for up to 34,840 cases and
211 114,981 controls from 38 genetic association studies with a case-control or cohort design.¹⁵
212 Fifty percent of the participants were women and the average age was 55 years.¹⁵ Imputation
213 was performed using the HapMap reference panel.¹⁵ Participant exclusion criteria
214 encompassed duplicate samples, relatedness, mismatch between self-reported and genotype-
215 determined sex, outlying heterozygosity and non-European descent. Type 2 diabetes status
216 was available in all participants. Data collection took place between 2002 and 2012.
217 Participant characteristics are reported in **Table 1** and further characteristics of studies
218 included in the DIAGRAM meta-analysis were reported previously in detail.¹⁵

219 The likelihood of bias for studies participating in this meta-analysis was deemed low on
220 the basis of: (a) the low proportion of participants with missing data on exposure or outcome,
221 (b) the high-quality genotyping or imputation of genetic variants included in the study
222 (**eTable 2**), (c) the low likelihood of bias by case-status in genotyping errors or genotype
223 misclassification, (d) the consideration that if any non-differential misclassification of
224 exposure or outcome occurred, that would result in a bias towards the null and (e) the
225 consideration that genetic variants are less likely to be affected by confounding or reverse
226 causality.^{16,17} On this basis, studies were deemed suitable for pooling by meta-analysis.

227 For the genetic variants included in these analyses, LDL cholesterol association
228 estimates were obtained from genetic association results in up to 188,577 participants of the
229 Global Lipids Genetics Consortium.¹⁸ In addition to type 2 diabetes, the association of these
230 LDL-lowering alleles with coronary artery disease and continuous cardiometabolic traits was
231 also estimated in large meta-analyses of genome-wide association studies. For coronary
232 artery disease, data were from the CARDIoGRAMplusC4D Consortium meta-analysis
233 (60,801 cases and 123,504 controls).¹⁹ For glycaemic traits, including fasting glucose^{20,21}
234 (N=133,010), glucose two hours after an oral glucose challenge^{20,22} (N=42,854) and fasting

235 insulin levels^{20,21} (natural-logarithm transformed; N=108,557), data were from the MAGIC
236 Consortium.²⁰⁻²² For anthropometric traits, including body mass index (N=333,495) and
237 waist-to-hip ratio (N=224,047), data were from the GIANT consortium.^{23,24} For details, see
238 **eTable 1**.

239 In exploratory analyses, the burden of protein-truncating and “probably deleterious”
240 missense variants in *NPC1L1*, *HMGCR*, *PCSK9*, *ABCG5*, *ABCG8* and *LDLR* was estimated
241 from exome sequencing studies of 8,373 type 2 diabetes cases and 8,466 controls (AMP-T2D
242 Program; T2D-GENES Consortium, SIGMA T2D Consortium. 2016 May 26;
243 <http://www.type2diabetesgenetics.org/>).

244

245 *Selection of genetic variants*

246 The combined association of two LDL cholesterol lowering genetic variants near
247 *NPC1L1* with type 2 diabetes constituted the primary analysis of the study (**Table 2**). These
248 variants were identified as having distinct effects on LDL cholesterol levels in approximate
249 conditional analyses using the GCTA software^{25,26} (see methodology description below;
250 **eFigure 2**). In sensitivity analyses, the combined association of five LDL-lowering alleles
251 near *NPC1L1*, previously used to predict the efficacy of ezetimibe,¹¹ was also investigated
252 (**eTable 3**).

253 For comparison with *NPC1L1*, other LDL-lowering alleles in or near genes encoding
254 other current or prospective molecular targets of LDL-cholesterol lowering therapy (i.e.
255 *HMGCR*, *PCSK9*, *ABCG5/G8*, *LDLR*) were studied.¹¹ Three LDL-lowering polymorphisms
256 in or near *HMGCR*, previously demonstrated to mimic the efficacy and metabolic effects of
257 statins,^{5,11} were analyzed (**Table 2**). At the *ABCG5/G8* and *LDLR* loci, polymorphisms
258 previously used to investigate genetic relationships between LDL cholesterol and coronary
259 artery disease¹¹ were studied (**Table 2**). At the *PCSK9* locus, in addition to the rs11591147

260 (p.R46L) variant (**Table 2**), the combined association of up to an additional eight likely-
261 independent LDL-lowering polymorphisms was investigated (**eFigure 3**). Genetic variants
262 included in the analyses were strongly and specifically associated with LDL cholesterol
263 (**eFigure 4**).

264 Approximate conditional analyses on large-scale LDL-cholesterol association data from
265 the Global Lipids Genetics Consortium¹⁸ using the GCTA software^{25,26} were performed in
266 order to identify distinct association signals for LDL cholesterol at the *NPC1L1* and *PCSK9*
267 loci. This approach uses genetic association results in addition to the linkage disequilibrium
268 pattern in a reference population to estimate the association of genetic variants in a region
269 after accounting for one or more index genetic variants. In so doing, the software allows for
270 the identification of likely-independent association signals in a given region using result-level
271 data. At the *PCSK9* locus, in a smaller sample of individuals with individual-level genotypes,
272 formal conditional analyses of the association with LDL cholesterol of polymorphisms after
273 adjusting for rs11591147 genotype status were also conducted (**eFigure 3**).

274

275 *Genetic reference information*

276 HUGO Gene Nomenclature Committee²⁷ (URL: www.genenames.org) gene names for
277 the investigated genes were: HGNC:7898 (*NPC1L1*), HGNC:5006 (*HMGCR*), HGNC:20001
278 (*PCSK9*), HGNC:13886 (*ABCG5*), HGNC:13887 (*ABCG8*), HGNC:6547 (*LDLR*). Genomic
279 coordinates reported in the manuscript represent the chromosome and physical position of
280 genetic variants according to the Human Reference Genome Build 37 (URL:
281 <http://www.ncbi.nlm.nih.gov/projects/genome/assembly/grc/>). Polymorphism names reported
282 in the manuscript represent rsID entries from dbSNP release 147 (URL:
283 <http://www.ncbi.nlm.nih.gov/SNP/>).

284

285 *Statistical analysis*

286 Genetic association data for the meta-analyses were either generated or gathered from
287 available sources at the MRC Epidemiology Unit, University of Cambridge (United
288 Kingdom). For each genetic variant and outcome, inverse variance weighted meta-analyses
289 using fixed-effect models was used to obtain pooled estimates. The I-squared statistic was
290 used to quantify heterogeneity. For each gene, associations of LDL-lowering genetic variants
291 and outcomes was estimated using Mendelian randomization statistical methodology.¹⁷
292 Estimates of “genetic variant to LDL-cholesterol” and “genetic variant to outcome”
293 associations were used to calculate estimates of “LDL-cholesterol reduction to outcome”
294 association at each gene.¹⁷ When multiple genetic variants at a given gene were included in
295 the model, estimates were pooled with a weighted generalized linear regression method that
296 accounts for the correlation between genetic variants.¹⁷ The correlation values were obtained
297 from the SNAP software²⁸ or from the 1000 Genomes Project data on individuals of
298 European ancestry (URL:<http://browser.1000genomes.org/>; **eTable 4**). Results were scaled to
299 represent the odds ratio per 1 mmol/L genetically-predicted reduction in LDL cholesterol.
300 Absolute risk differences were estimated assuming that the incidence rate of type 2 diabetes
301 in the InterAct study subcohort would be the baseline incidence rate in “non-exposed”
302 individuals (i.e. 3.76 incident cases per 1000 person-years of follow-up).¹³ This baseline rate
303 was then multiplied by the odds ratio estimated from genetic analyses to obtain the “at risk”
304 incidence rate. The absolute risk difference estimate was the “at risk” incidence rate minus
305 the baseline incidence rate. Absolute risk differences were expressed in incident events per
306 1000 person-years for a 1 mmol/L genetically-predicted reduction in LDL cholesterol.
307 Statistical analyses were conducted using STATA v14.1 (StataCorp, College Station, Texas
308 77845 USA), R v3.2.2 (The R Foundation for Statistical Computing), and METAL.²⁹ A two-
309 tailed p-value of $p < 0.05$ was considered statistically significant.

310 **Results**

311 *LDL cholesterol lowering alleles at NPC1L1 and risk of type 2 diabetes*

312 LDL cholesterol lowering alleles at the *NPC1L1* locus were inversely associated with
313 coronary artery disease and directly associated with type 2 diabetes, both individually (**Table**
314 **2**) and collectively (odds ratio of type 2 diabetes per 1 mmol/L genetically-predicted LDL
315 cholesterol reduction, 2.42; 95% confidence interval, 1.70-3.43, $p < 0.001$; estimated absolute
316 risk difference, 5.3 incident cases per 1000 person-years for a 1 mmol/L genetically-predicted
317 reduction in LDL cholesterol; **Figure 1**). For both polymorphisms, estimates of the
318 association with type 2 diabetes were consistent across the studies included in the meta-
319 analysis (**eFigure 1**). In the periphery of the *NPC1L1* locus, approximately 400 kilobases
320 from the lead rs2073547 polymorphism, there was a known association signal for type 2
321 diabetes and glycemic traits near the *GCK* gene.^{15,20,21} After accounting for variation in *GCK*,
322 the association of with type 2 diabetes at *NPC1L1* did not change (**eTable 3**). Association
323 estimates also remained unchanged when modeling the association of five polymorphisms
324 previously used by Ference et al.¹¹ as a proxy for *NPC1L1* inhibition (**eTable 3**). In 14,657
325 cases of type 2 diabetes and 118,854 controls for whom we had access to individual-level
326 genotyping data, there was no evidence of heterogeneity in the association between *NPC1L1*
327 alleles and type 2 diabetes in analyses stratified by age, sex or body mass index (**eFigure 5**).
328 In exome sequencing association results, there was no evidence of enrichment of *NPC1L1*
329 protein truncating alleles in type 2 diabetes cases compared with controls (odds ratio of type
330 2 diabetes for individuals carrying a truncating allele, 1.12; 95% confidence interval, 0.88-
331 1.43; $p = 0.34$), but missense variants in *NPC1L1* predicted to be “probably deleterious” were
332 overrepresented in individuals with type 2 diabetes compared with controls (1.26, 1.07-1.47;
333 $p = 0.005$).

334

335

336 *Associations with type 2 diabetes at other genes*

337 As previously reported,^{5,11} LDL cholesterol lowering alleles at *HMGCR* were also
338 associated with type 2 diabetes and coronary artery disease in opposite directions (**Table 2**
339 **and Figure 1**). An association of the loss-of-function p.R46L (rs11591147) variant in *PCSK9*
340 with higher risk of type 2 diabetes was also found (odds ratio of type 2 diabetes per 1 mmol/L
341 genetically-predicted LDL cholesterol reduction, 1.19; 95% confidence interval, 1.02-1.38,
342 p=0.03; estimated absolute risk difference, 0.7 incident cases per 1000 person-years for a 1
343 mmol/L genetically-predicted reduction in LDL cholesterol; **Table 2 and Figure 1**). At
344 *PCSK9*, analyses of the LDL cholesterol association data suggested the presence of distinct
345 association signals. In formal conditional analyses, there was evidence of at least two distinct
346 association signals (rs11591147 and rs471705; **eFigure 3**). Using the GCTA software,^{25,26}
347 approximate conditional analyses suggested the presence of nine distinct association signals
348 (rs11591147 plus eight additional genetic variants; **eFigure 3**). Inclusion of these additional
349 signals gave similar associations with type 2 diabetes as the p.R46L variant alone (odds ratio
350 of type 2 diabetes per 1 mmol/L genetically-predicted reduction in LDL cholesterol using
351 rs11591147 plus rs471705, 1.21, 95% confidence interval, 1.04-1.41, p=0.01; and 1.16, 1.03-
352 1.31, p=0.02, using rs11591147 plus the eight additional polymorphisms; **eTable 3**). The
353 association with type 2 diabetes of LDL-lowering alleles at the *ABCG5/G8* and *LDLR* loci
354 did not reach statistical significance. There was no evidence of association with type 2
355 diabetes for missense variants predicted to be “probably deleterious” or protein truncating
356 alleles in the *HMGCR*, *PCSK9*, *ABCG5*, *ABCG8* and *LDLR* genes (**eTable 5**), but the
357 confidence intervals around risk estimates were generally wide, reflecting the low prevalence
358 of these genetic variants and the relatively small sample size of this analysis.

359

360

361 *Evidence of gene-specific associations with type 2 diabetes risk*

362 In analyses of the association with disease risk for a given genetically-predicted
363 reduction in LDL cholesterol, there was a similar reduction in coronary artery disease risk
364 across genes (I-squared for heterogeneity in genetic associations = 0.0%; p=0.93, **Figure 1**).
365 However, for the same reduction in LDL cholesterol, the association with type 2 diabetes risk
366 differed by gene (I-squared = 77.2%; p=0.002, **Figure 1**). The different magnitudes and
367 directions of association of LDL-lowering alleles with continuous glycemc and
368 anthropometric traits suggested gene-specific mechanisms underlying the altered risk of type
369 2 diabetes (**eFigure 6**). For example, at the *HMGCR* locus there were associations with body
370 mass index and waist-to-hip ratio, while at the *PCSK9* locus there were associations with
371 higher fasting glucose and two hour glucose (**eFigure 6**).

372 **Discussion**

373 In this meta-analysis, exposure to LDL-cholesterol lowering genetic variants in or near
374 the *NPC1L1* gene was associated with a higher risk of type 2 diabetes. This finding is
375 consistent with the results of a small-scale open label randomized clinical trial, showing
376 increased glycated hemoglobin in association with ezetimibe treatment.³⁰ Blazing et al.
377 reported that the addition of ezetimibe to simvastatin for secondary cardiovascular prevention
378 in the IMPROVE-IT trial resulted in a small and not statistically significant increase in risk of
379 new-onset diabetes (9% relative risk increase per 1 mmol/L reduction in LDL cholesterol).³¹
380 However, IMPROVE-IT results may not be sufficient to rule out an effect of inhibiting
381 Niemann-Pick C1-like 1 on diabetes risk because: (1) some of the effects of NPC1L1
382 inhibition may be apparent only after several years of treatment; (2) the risk of type 2
383 diabetes in individuals with a history of acute coronary syndrome yet free from type 2
384 diabetes in IMPROVE-IT may not reflect that of the general population on which this genetic
385 analysis is based; (3) limited compliance to drug treatment, as observed in IMPROVE-IT,⁷
386 may dilute etiologic effect estimates. By analogy, the association of statin treatment with
387 higher diabetes risk was only demonstrable in a meta-analysis of several randomized clinical
388 trials including more than 90,000 individuals.³ Therefore, these results warrant the continued
389 monitoring of the glycemic effects of ezetimibe in randomized clinical trials and clinical
390 practice particularly in a primary prevention setting.

391 The results of this study show that multiple LDL-lowering mechanisms, including those
392 mediated by the molecular targets of available LDL-lowering drugs (i.e. statins, ezetimibe,
393 and PCSK9 inhibitors), are associated with adverse metabolic consequences and increased
394 type 2 diabetes risk. These findings are consistent with other studies of the association with
395 type 2 diabetes of genetic scores aggregating multiple polymorphisms affecting LDL
396 cholesterol and other lipid fractions.³² They are also consistent with the observation that

397 patients with familial hypercholesterolemia are less likely to have type 2 diabetes.³³ The
398 genes which were associated both with lower LDL cholesterol levels and higher type 2
399 diabetes risk impact on LDL cholesterol by distinct pathways including cholesterol
400 absorption (*NPC1L1*),³⁴ endogenous cholesterol synthesis (*HMGCR*)³⁵ and internalization of
401 cholesterol-rich particles into the cell (*PCSK9*).^{36,37} For a similar reduction in LDL
402 cholesterol, the association with type 2 diabetes differed by gene which would be consistent
403 with the mediation of their associations by different mechanisms. Besseling et al. have
404 proposed that an increased internalization of cholesterol into pancreatic beta-cells may result
405 in impaired secretion of insulin,³³ a hypothesis supported by murine experimental models.³⁸
406 LDL cholesterol lowering alleles at *HMGCR* are associated with higher fasting insulin and
407 body mass index, suggesting an insulin resistance-related mechanism.⁵ Finally, in contrast
408 with early evidence showing metabolic benefits of *NPC1L1* knock-out in mice,³⁹ recent
409 studies suggest that its over expression in the liver may suppress gluconeogenesis and,
410 therefore, that its inhibition could perhaps enhance glucose production.⁴⁰ Overall, these
411 results indicate complex relationships between the mechanisms leading to lower LDL
412 cholesterol and metabolic risk.

413 Contrary to previous, smaller-scale investigations,⁴¹ there were associations of the
414 p.R46L variant in *PCSK9* (rs11591147) with a higher risk of type 2 diabetes, and higher
415 fasting and two hour glucose. These associations have to be interpreted with caution, given
416 the level of statistical significance for the association and the context of multiple comparisons
417 presented in this study. This finding nonetheless suggests that the effect of LDL-lowering
418 drugs on increased diabetes risk might extend to the newly-developed PCSK9 inhibitors,
419 encouraging further genetic and clinical trial investigations.

420 In general, unlike the association of LDL-lowering alleles with cardiovascular risk, the
421 association of these alleles with metabolic risk appears to be gene-specific, which in turn

422 might suggest that the adverse consequences of lipid-lowering agents on diabetes risk could
423 be target-specific. This may have clinical implications for the future of lipid-lowering therapy
424 in the context of the growing number of approved drugs acting on different molecular targets.
425 The overall safety profile of these drugs, including the magnitude of risk of new-onset type 2
426 diabetes, may be relevant to the choice of specific agent for subsets of the patient population,
427 for example those at high risk of type 2 diabetes who are also candidates for lipid-lowering
428 therapy.

429 A number of assumptions and potential limitations of the genetic approach used in this
430 study should be considered. “Mendelian randomization” generally assumes that genetic
431 variants are associated with the endpoint exclusively via the risk factor of interest.^{16,17} The
432 strong and specific association with LDL cholesterol, the well-known role of target genes in
433 LDL cholesterol metabolism and the use of conditionally-distinct genetic variants at given
434 loci strengthen the validity of the genetic models used in this study. Similar to previous
435 examples,^{5,11,42} the aim of this study was to use genetic variants that “mimic” the action of
436 pharmacological therapy and therefore “pleiotropy” (i.e. the association with variables other
437 than LDL cholesterol) may be more informative than concerning. For instance, *HMGCR*
438 genetic variants are associated with higher body mass index, consistent with the effects on
439 body weight observed in randomized clinical trials of statins.⁵ However, the consequences of
440 modest reductions in LDL cholesterol associated with LDL-lowering alleles over several
441 decades, as assessed in this study, may differ from the short-term pharmacological inhibition
442 of a molecular target in randomized clinical trials or clinical practice. Finally, several of the
443 included studies were population-based and therefore association estimates from these studies
444 may not be applicable to patient groups in whom a particular therapy is indicated.

445

446

447 **Conclusions**

448 In this meta-analysis, exposure to LDL-cholesterol lowering genetic variants in or
449 near *NPC1L1* and other genes was associated with a higher risk of type 2 diabetes. These data
450 provide insights into potential adverse effects of LDL cholesterol-lowering therapy.

451

452 **Table 1. Participants of EPIC-InterAct, UK Biobank and DIAGRAM.**

453

Variable	EPIC-InterAct		UK Biobank		DIAGRAM	
	Type 2 diabetes	Subcohort (non-cases)	Type 2 diabetes	Controls	Type 2 diabetes	Controls
Country	Multiple European countries		United Kingdom		Europe and United States ^c	
Genotyping chip	Illumina 660w quad and Illumina CoreExome chip		Affymetrix UK Biobank Axiom Array		Multiple ^d	
Imputation panel	Haplotype Reference Consortium		1000 Genomes Phase 3 plus UK10K		HapMap	
Number	10,071 ^a	12,423 ^a	6,627	143,765	34,840	114,981
Age, mean years (SD)	56 (8)	52 (9)	60 (7)	56 (8)	59 (10)	54 (14)
Female sex, N (%)	5,037 (50)	7,713 (62)	2,349 (35)	77,397 (54)	14,621 (42)	60,377 (53)
Smoking status, current smokers N (%)	2,830 (28)	3,240 (26)	811 (12)	18,149 (13)	NA	NA
BMI in kg/m ² , mean (SD)	29.7 (4.8)	25.8 (4.1)	31.9 (5.9)	27.3 (4.7)	29.7 (5.9)	26.5 (4.5)
Waist-to-hip ratio	0.92 (0.09)	0.85 (0.09)	0.95 (0.08)	0.87 (0.09)	NA	NA
Systolic blood pressure in mmHg, mean (SD)	144 (20)	132 (19)	141 (17)	138 (19)	NA	NA
Diastolic blood pressure in mmHg, mean (SD)	87 (11)	82 (11)	82 (10)	82 (10)	NA	NA
LDL cholesterol in mmol/L, mean (SD)	4.0 (1)	3.8 (1)	NA ^b	NA ^b	NA	NA
HDL cholesterol in mmol/L, mean (SD)	1.2 (0.4)	1.5 (0.4)	NA ^b	NA ^b	NA	NA
Triglycerides in mmol/L, median (IQR)	1.7 (1.2-2.5)	1.1 (0.8-1.6)	NA ^b	NA ^b	NA	NA

454 Abbreviations: N, number of participants; BMI, body mass index; SD, standard deviation; LDL, low-density lipoprotein
 455 cholesterol; HDL, high-density lipoprotein cholesterol; IQR, interquartile range; NA, not available.

456 a A total of 9,308 type 2 diabetes cases and 11,523 non-cases were included in the main analysis of the association of genetic
 457 variants with type 2 diabetes after the exclusion of participants overlapping with DIAGRAM.

458 b Blood lipids concentrations are being measured in the UK Biobank study, with data release currently planned for the end
 459 of 2016.

460 c DIAGRAM had a small South Asian component accounting for 2.44% of participants.

461 d Affymetrix Human SNP Array 6.0; Illumina HumanHap 300, 300/370 and 550; Affymetrix Genechip 500K & MIPS 50K;
 462 Cardio-Metabolchip.

463

464 **Table 2. LDL-cholesterol lowering polymorphisms at *NPC1L1* and other genes and their association with type 2 diabetes and coronary**
 465 **artery disease.**
 466

Gene	dbSNP rsID	Genomic coordinate, chromosome and position	Effect / other allele	Effect allele frequency, mean (range)	LDL, N	Beta (95% CI) per allele in mmol/L ^a	p-value	OR of coronary artery disease (95% CI) per allele ^b	p-value	OR of type 2 diabetes (95% CI) per allele ^c	p-value	I-squared	Heterogeneity p-value
<i>NPC1L1</i>	rs2073547	chr7:44582331	A / G	0.81 (0.81, 0.82)	169,889	-0.049 (-0.058, -0.039)	2 x 10 ⁻²¹	0.980 (0.957, 1.003)	0.09	1.051 (1.027, 1.075)	2 x 10 ⁻⁰⁵	0.0%	0.48
	rs217386	chr7:44600695	A / G	0.42 (0.41, 0.44)	173,021	-0.036 (-0.044, -0.029)	1 x 10 ⁻¹⁹	0.979 (0.960, 0.998)	0.03	1.027 (1.009, 1.045)	0.003	0.0%	0.68
<i>HMGCR</i>	rs12916 ^d	chr5:74656539	T / C	0.58 (0.57, 0.60)	168,357	-0.073 (-0.081, -0.066)	8 x 10 ⁻⁷⁸	0.965 (0.947, 0.983)	0.0002	1.029 (1.012, 1.046)	0.0007	46.4%	0.03
	rs5744707	chr5:74890618	A / G	0.90 (0.90, 0.91)	172,928	-0.055 (-0.067, -0.043)	6 x 10 ⁻¹⁹	0.970 (0.941, 0.999)	0.04	0.983 (0.956, 1.011)	0.24	2.8%	0.38
	rs16872526	chr5:74675717	T / G	0.91 (0.90, 0.92)	173,009	-0.041 (-0.054, -0.027)	2 x 10 ⁻⁰⁸	0.988 (0.959, 1.018)	0.44	1.016 (0.985, 1.047)	0.32	50.1%	0.11
<i>PCSK9</i>	rs11591147	chr1:55505647	T / G	0.02 (0.01, 0.02)	77,417	-0.497 (-0.532, -0.462)	9 x 10 ⁻¹⁴³	0.774 (0.692, 0.866)	7 x 10 ⁻⁰⁶	1.089 (1.010, 1.174)	0.03	0.0%	0.39
<i>ABCG5/G8</i>	rs4299376	chr2:44072576	T / G	0.69 (0.68, 0.70)	144,861	-0.081 (-0.090, -0.072)	4 x 10 ⁻⁷²	0.950 (0.931, 0.970)	1 x 10 ⁻⁰⁶	1.011 (0.990, 1.032)	0.29	2.2%	0.38
<i>LDLR</i>	rs6511720	chr19:11202306	T / G	0.11 (0.10, 0.12)	170,608	-0.221 (-0.233, -0.209)	4 x 10 ⁻²⁶²	0.882 (0.853, 0.912)	1 x 10 ⁻¹³	1.028 (0.999, 1.057)	0.05	0.0%	0.93

467 Polymorphism names reported in the table are rsID entries from dbSNP release 147.

468 Genomic coordinates represent chromosome and physical position of genetic variants according to the Human Reference Genome Build 37.

469 Effect allele frequency averages and ranges are from EPIC-InterAct,¹³ UK Biobank¹⁴ and DIAGRAM.¹⁵

470 Abbreviations: LDL, low-density lipoprotein cholesterol; N, number of participants; CI, confidence interval; CAD, coronary artery disease; T2D, type 2 diabetes; OR, odds ratio.

471 a LDL cholesterol data were from the Global Lipids Genetics Consortium.¹⁸

472 b Coronary artery disease data were from 60,801 coronary artery disease cases and 123,504 controls from the CARDIoGRAMplusC4D Consortium.¹⁹

473 c Type 2 diabetes data were from 50,775 cases of type 2 diabetes and 270,269 controls from EPIC-InterAct,¹³ UK Biobank¹⁴ and DIAGRAM.¹⁵

474 d In addition to EPIC-InterAct,¹³ UK Biobank¹⁴ and DIAGRAM.¹⁵, type 2 diabetes association analyses of rs12916 included eleven studies (4,496 cases and 50,677 controls)
 475 previously reported by Swerdlow and colleagues.⁵ The total sample size of this analysis was of 55,271 cases of type 2 diabetes and 320,946 controls.
 476

477 **Figure Legend**

478 **Figure 1 Title. Odds ratio of coronary artery disease and type 2 diabetes associated with**
479 **LDL-lowering genetic variants in or near investigated genes.**

480

481 **Figure 1 Footnote:** Coronary artery disease data were from 60,801 coronary artery disease cases and 123,504
482 controls from the CARDIoGRAMplusC4D Consortium.¹⁹ Type 2 diabetes data were from 50,775 cases of type
483 2 diabetes and 270,269 controls from EPIC-InterAct,¹³ UK Biobank¹⁴ and DIAGRAM.¹⁵ In addition to EPIC-
484 InterAct,¹³ UK Biobank¹⁴ and DIAGRAM,¹⁵ type 2 diabetes association analyses of rs12916 at *HMGCR*
485 included eleven studies (4,496 cases and 50,677 controls) previously reported by Swerdlow and colleagues.⁵
486 Therefore the sample size of *HMGCR* genetic variants association with type 2 diabetes was of up to 55,271
487 cases of type 2 diabetes and 320,946 controls. All results are scaled to represent the odds ratio per 1 mmol/L
488 genetically-predicted reduction in LDL cholesterol. Abbreviations: SNP, single nucleotide polymorphism; OR,
489 odds ratio; CAD, coronary artery disease; T2D, type 2 diabetes; LDL, low-density lipoprotein cholesterol.

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522 **References**

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Figure 1

