

1 Coronary artery disease risk and lipidomic 2 profiles are similar in familial and 3 population-ascertained hyperlipidemias

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25 **Abstract**

26 **Aims:** To characterize and compare coronary artery disease (CAD) risk and detailed lipidomic profiles
27 of individuals with familial and population-ascertained hyperlipidemias.

28 **Methods and Results:** We determined incident CAD risk for 760 members of 66 hyperlipidemic
29 families (≥ 2 first degree relatives with the same hyperlipidemia) and 19,644 Finnish FINRISK
30 population study participants. We also quantified 151 lipid species in plasma or serum samples from
31 550 members of 73 hyperlipidemic pedigrees and 897 FINRISK participants using a mass spectrometric
32 shotgun lipidomics platform. Hyperlipidemias (LDL-C or triacylglycerides over 90th population
33 percentile) were associated with increased CAD risk (high LDL-C: HR 1.74, 95% CI 1.48–2.04; high
34 triacylglycerides: HR 1.38, 95% CI 1.09–1.74) and the risk estimates were very similar between the
35 family and population samples. High LDL-C was associated with altered levels of 105 lipid species in
36 families (p -value range 0.033– 7.3×10^{-20} at 5% false discovery rate) and 51 species in the population
37 samples (p -value range 0.017– 6.8×10^{-21}). Hypertriglyceridemia was associated with altered levels of
38 117 lipid species in families (p -value range 0.035– 1.8×10^{-49}) and 119 species in the population sample
39 (p -value range 0.038– 2.3×10^{-56}). The lipidomics profiles of hyperlipidemias were highly similar in
40 families and population samples.

41 **Conclusion:** We identified distinct lipidomic profiles associated with high LDL-C and triacylglyceride
42 levels. CAD risk, lipidomic profiles and genetic profiles are highly similar between familial and
43 population-ascertained hyperlipidemias, providing evidence of similar and overlapping underlying
44 mechanisms. Our results do not support different screening and treatment for such hyperlipidemias.

45 **Keywords:** coronary artery disease; hyperlipidemia; low-density lipoprotein cholesterol; triglycerides,
46 lipids; family study

47 **Introduction**

48 High levels of low-density lipoprotein cholesterol (LDL-C) and triacylglycerides (TGs) have been
49 identified as causal risk factors for atherosclerotic cardiovascular disease (ASCVD).^{1, 2} These
50 hyperlipidemias may arise through lifestyle factors, but they are also highly heritable.³⁻⁶ More than half
51 of patients with premature coronary artery disease (CAD) may be affected by familial dyslipidemias, a
52 majority of which are characterized by elevations in LDL-C and/or TGs.⁷

53 Whether familial hyperlipidemias should be diagnosed and managed differently from
54 hyperlipidemias observed in randomly ascertained individuals in the general population is uncertain.
55 Carriers of rare high-impact variants predisposing to familial hypercholesterolemia (FH) may have a
56 higher risk of developing coronary artery disease than non-carriers with similar lipid levels.⁸ This is
57 potentially related to lifelong exposure to high LDL-C levels and suggests that these individuals may
58 benefit from earlier or more aggressive LDL-C-lowering therapy.

59 Rare variants, however, explain only a small fraction of familial hyperlipidemias with either
60 elevated LDL-C or TG levels. Instead, there is growing evidence that familial hyperlipidemias are highly
61 polygenic in origin.⁸⁻¹³ It is of high clinical importance whether non-monogenic familial hyperlipidemias
62 represent distinct disease entities and lead to different ASCVD susceptibility compared with population-
63 ascertained hyperlipidemias.¹⁴ A recent study provided suggestive evidence to the contrary, finding
64 greater preclinical atherosclerosis in monogenic FH than in polygenic hypercholesterolemia.¹⁵ However,
65 it is unknown how these results translate to clinical outcomes between familial and population-
66 ascertained hyperlipidemias.

67 How the heterogenic genetic background of familial hyperlipidemias manifests in detailed
68 circulating lipid profiles, and reflects underlying pathophysiology, is also unknown. Recent
69 technological advancements have allowed replicable and simultaneous quantification of hundreds of
70 lipid species through lipidomic profiling.^{16, 17} We tested whether precise phenotypic differences in
71 lipidomic profiles, which might underlie differing ASCVD susceptibility, exist between familial and

72 population-ascertained hyperlipidemias after excluding individuals with monogenic FH. We looked
73 beyond the traditional clinical measures of LDL-C and TG concentrations at the numerous fatty acid
74 ester species of TGs and cholesterol species which are the major constituents of LDL and TG-rich
75 lipoprotein particles. We used a direct infusion platform that combines absolute quantification with high
76 throughput, overcoming problems that have hampered many previous studies.¹⁸

77 Lipid species including sphingolipids, glycerophospholipids, glycerolipids and cholesteryl
78 esters may predict ASCVD incidence or event risk over traditional risk factors, and the levels of many
79 individual lipid species have been linked to distinct genetic loci.¹⁹⁻²² Major differences in the metabolic
80 pathways underlying different types of hyperlipidemias would thus be expected to be reflected in
81 different lipidomic profiles. As an example, individuals with low HDL-C levels have previously been
82 shown to have low phosphatidylethanolamine-plasmalogen levels in HDL particles, a marker of HDL
83 anti-oxidative capacity.²³

84 In the present study, we first estimated the CAD risk associated with familial and population-
85 ascertained hyperlipidemias. Secondly, we characterized the lipidomic profiles associated with elevated
86 plasma levels of LDL-C and TGs. Finally, we compared the lipidomic profiles of familial and
87 population-ascertained hyperlipidemias to assess their potential differences.

88 **Materials and Methods**

89 **Subjects and clinical ascertainment**

90 The Finnish hyperlipidemia families included in this cohort study (74 families, $n = 1,445$ individuals
91 with at least LDL-C and TG measures) were identified as part of The European Multicenter Study on
92 Familial Dyslipidemias in Patients with Premature Coronary Heart Disease (EUFAM). Designation of
93 “familial high LDL-C” or “familial high TGs” was made if at least two first-degree relatives of each
94 other had LDL-C or TG levels, respectively, that were $> 90^{\text{th}}$ age- and sex-specific Finnish 1997
95 population percentiles (Supplemental Table IV). Classic familial hypercholesterolemia (FH) was
96 excluded based on genotyping and an in-house functional low-density lipoprotein receptor test.

97 Samples from the Finnish National FINRISK study were used as a Finnish population-based
98 comparison group. 19,644 individuals from the FINRISK 1992-2002 cohorts and 755 individuals from
99 EUFAM families passed exclusion criteria and could be linked with the national hospital discharge and
100 causes-of-death registries. Clinical incident CAD event endpoints were defined as either myocardial
101 infarction or coronary revascularization (coronary angioplasty or coronary artery bypass grafting). Mean
102 (range) follow-up time from baseline to CAD endpoint, death, or end of registry follow-up was 16.1
103 (0.1–20.1) years in EUFAM and 12.6 (0.02–19.0) years in FINRISK. More detailed information is given
104 in the Supplementary material online.

105 **Lipidomics measurements**

106 Lipidomic profiling and analysis of circulating lipid species was performed for 550 EUFAM family
107 members with available plasma samples and for 897 individuals from the FINRISK 2012 cohort. Mass
108 spectrometry-based lipid analysis was performed at Lipotype GmbH (Dresden, Germany) as
109 described.¹⁶ Plasma and serum lipids were extracted with methyl tert-butyl ether/methanol (7:2, V:V)
110 as in Matyash et al.²⁴ Samples were analyzed by direct infusion in a QExactive mass spectrometer
111 (Thermo Scientific) equipped with a TriVersa NanoMate ion source (Advion Biosciences). Samples
112 were analyzed in both positive and negative ion modes in a single acquisition.

113 Data were analyzed with in-house developed lipid identification software based on
114 LipidXplorer.^{25, 26} Reproducibility was assessed by the inclusion of reference plasma samples. Median
115 coefficient of variation was <10% across all batches. A total of 151 species were detected in $\geq 80\%$ of
116 both EUFAM and FINRISK samples and were included in the subsequent analyses. Right-skewed
117 lipidomics measures were natural logarithm transformed prior to normalization by standard deviation.
118 More detailed information is given in the Supplementary material online.

119 **Statistical analyses**

120 To assess the risk of incident coronary artery disease associated with the hyperlipidemias, we used Cox
121 proportional hazards models using age as the time scale, stratified by sex, and clustered by family to
122 estimate hazard ratios (HR) for incident CAD events, excluding individuals with prevalent CAD.

123 We used linear mixed models to estimate the association between lipidomic parameters and
124 predictors of interest (hyperlipidemia status, continuous lipid measurement, or genotype) as
125 implemented in MMM (version 1.01).²⁷ Age, age², and sex were used as additional fixed effect
126 covariates. To account for relatedness among individuals, an empirical genetic relationship matrix was
127 included as the covariance structure of a random effect. Statistical significance was evaluated using the
128 Benjamini-Hochberg method at the 5% level to account for multiple comparisons. R (version 3.4.3) was
129 used for data transformations and other analyses.²⁸ Detailed information is given in the Supplementary
130 material online.

131 **Methods**

132 **Clinical characteristics and CAD risk of individuals with high levels of LDL-C or TGs**

133 We first assessed the risk of developing CAD associated with high levels of LDL-C or TGs in
134 individuals from the Finnish FINRISK population survey and in hyperlipidemic pedigrees ascertained
135 as part of the European Study of Familial Hyperlipidemias (EUFAM) (Table 1; Supplemental Table I;
136 Supplemental Figure I.A.). Individuals with high LDL-C had an increased risk of being diagnosed with
137 CAD in the FINRISK population surveys ($n = 19,644$ individuals) compared to other individuals (HR
138 1.74; 95% CI 1.48–2.05). The members of hyperlipidemic families with high LDL-C had a similar
139 though nonsignificant HR compared to their relatives without high LDL-C in 47 “high LDL-C” families
140 ($n = 633$ individuals) (HR 1.71; 95% CI 0.94–3.10). The HRs did not differ between the cohorts
141 ($p=0.60$). We also observed an increased CAD risk in individuals with high TGs in the population (HR
142 1.38; 95% CI 1.08–1.75), and a similar though nonsignificant HR in 35 “high TG” families ($n = 375$
143 individuals) (HR 1.35; 95% CI 0.52–3.51). The HRs did not differ between the cohorts ($p=0.91$). Meta-
144 analyses of HRs closely approximated estimates derived in the population cohort.

145 *Position of Table 1*

146 We then characterized the detailed lipidomic profiles of 550 individuals from 73 hyperlipidemic

147 families and 897 individuals from the FINRISK population study (Methods; Supplemental Table II;
148 Supplemental Figure I.B.). These included 105 individuals (23 %) out of 463 family members in 53
149 “high LDL-C” families who had LDL-C levels >90th percentile (mean \pm SE 5.2 ± 0.8 mmol/l), and 64
150 individuals (22 %) out of 287 family members in 39 “high TG” families who had TGs >90th percentile
151 (3.6 ± 2.0 mmol/l). Using similar cutoffs in the population, 56 individuals (6 %) and 65 individuals (7
152 %) out of 897 individuals were affected by high LDL-C levels (5.3 ± 1.1 mmol/l) and high TGs ($3.5 \pm$
153 1.6 mmol/l) respectively. Simultaneously high LDL-C and TG levels were observed in 31 individuals
154 in the family samples and 9 individuals in the population samples.

155 **High LDL-C and lipidomic profiles**

156 To characterize the lipidomic profiles associated with elevated values of LDL-C, we compared
157 individuals with high LDL-C levels vs. those without. In the hyperlipidemic families, individuals with
158 high LDL-C had significantly elevated levels of 99 lipid species spread out across most of the studied
159 lipid classes. Reduced levels were observed for three lysophosphatidylcholine (LPC), two
160 lysophosphatidylethanolamine (LPE) and one phosphatidylcholine-ether (PCO) species (Figure 1.A.;
161 Supplemental Table III). Similar trends were seen in the population when comparing individuals with
162 high LDL-C vs. those without. We observed significantly elevated levels of 51 lipid species (Figure
163 1.B.; Supplemental Table III). The effect estimates correlated strongly across all lipid species between
164 the hyperlipidemic families and the population cohorts (Pearson’s $r = 0.80$; Figure 3). Further, we
165 observed no statistically significant cohort-dyslipidemia interactions at the 5% false discovery rate.

166 *Position of Figure 1*

167 We also studied the association of high LDL-C levels with the degree of saturation of fatty acids
168 in each lipid class. In the hyperlipidemic families, high LDL-C levels were associated with increased
169 saturation of LPCs and ceramides, as well as reduced saturation of LPEs, PCs, phosphatidylcholine-
170 ethers (PCOs), and phosphatidylinositols (PIs) (p -value range = 0.019–0.0014) (Supplemental Figure
171 II). In the population samples, the trends were similar, although there was an association for increased
172 LPC saturation only ($p=7.2*10^{-4}$). There were no cohort-hyperlipidemia interactions in any of the lipid

173 species at the 5% false discovery rate. Overall, the lipidomic profiles associated with high LDL-C levels
174 appeared similar in the hyperlipidemic families and the general population.

175 **High TGs and lipidomic profiles**

176 In the hyperlipidemic families, individuals with high TGs had elevated levels of 107 lipid species
177 covering all studied lipid classes with the exception of LPEs. In addition, we observed reduced levels
178 of seven PCO, two LPC, and one PI species (Figure 2.A.; Supplemental Table III). Similar profiles were
179 seen in the population when comparing individuals with high TGs vs. those without. We observed
180 elevated levels of 108 species and reduced levels of ten PCO and one LPC species (Figure 2.B.;
181 Supplemental Table III). The effect estimates correlated very highly across all species between families
182 and population samples (Pearson's $r = 0.96$; Figure 3). Further, we observed no cohort-dyslipidemia
183 interactions for any of the lipid species at the 5% false discovery rate.

184 *Position of Figure 2*

185 *Position of Figure 3*

186 Next, we studied the association of high TG levels with the degree of saturation of fatty acids in each
187 lipid class. In both the hyperlipidemic families and the population, having high TGs was significantly
188 associated with increased saturation of TGs, DGs, LPCs, and CEs (p -value range = 0.0012 – 5.9×10^{-11})
189 (Supplemental Figure III). There were no statistically significant cohort-hyperlipidemia interactions at
190 the 5% false discovery rate. Overall, we observed great similarity in the lipidomic profiles associated
191 with high TGs in the hyperlipidemic families and in the general population.

192 **Independent associations of LDL-C and TG values with the lipid species**

193 Since we observed modest combined hyperlipidemia in both cohorts, we estimated the independent
194 associations of LDL-C and TGs with each lipid species in co-adjusted models (Figure 4; Supplemental
195 Table III). In these analyses, many of the observed associations with LDL-C were greatly diluted in
196 magnitude. LDL-C levels remained most strongly associated with CE, SM, ceramide, PC and PCO

197 species in both cohorts. A total of 83 species in the hyperlipidemic families and 91 species in the
198 population were independently associated with LDL-C at the 5% false discovery rate. In contrast, TGs
199 remained strongly associated with a wide range of lipid species, including all individual TG species,
200 DGs, PCs, PEs, PIs, ceramides and a subset of CEs in both cohorts. A total of 125 species in the
201 hyperlipidemic families and 124 species in the population were independently associated with TGs at
202 the 5% false discovery rate. Overall, only 13 species were uniquely associated with LDL-C in either
203 cohort, whereas 42 species were uniquely associated with TGs (Supplemental Figure IV).

204 *Position of Figure 4*

205 **Discussion**

206 Recent lipidomic approaches have identified several hundreds of different lipid species in the human
207 circulation, some of which could be better prognostic biomarkers for atherosclerotic cardiovascular
208 disease than the traditional clinical chemistry measurements. To the best of our knowledge, the present
209 study is the most comprehensive lipidomic profiling of common hyperlipidemias to date. We used a
210 mass spectrometric lipidomics platform to assess the lipidomic profiles in individuals with high LDL-C
211 and/or TG levels. We found that individuals affected by high levels of LDL-C or TGs had CAD HRs
212 between 1.34–1.74 in the family and population cohorts and exhibited distinct lipidomic profiles with
213 clear variation between lipid classes. In total, out of 151 lipidomic species, 108 were significantly
214 associated with high LDL-C and 131 with high TG levels in at least one cohort. Of these, 96 species
215 were common with both high LDL-C and TGs. In addition, we observed highly similar lipidomic
216 profiles between the familial and population-ascertained hyperlipidemias.

217 These findings allow us to draw several conclusions. First, the CAD risks are highly similar
218 regardless of whether hyperlipidemic individuals were identified from families with a high prevalence
219 of hyperlipidemia or from the general population. This result contrasts with earlier results reporting
220 much higher CAD risks in relatives of familial combined hyperlipidemia probands compared to
221 spouses.²⁹ Our study, however, compares the estimates between family members to individuals with

222 closely similar lipid levels from the population to isolate the effect of familiarity. We also study the risk
223 associated with elevated LDL-C and TGs separately. Our estimates are also lower than typically reported
224 for monogenic FH, which may be associated with lifelong exposure to very high LDL-C levels and
225 increased CAD risk compared with other individuals with similarly extreme hyperlipidemias.⁸ In the
226 present study, we excluded probands with monogenic FH based on a functional LDL receptor test.
227 Excepting monogenic FH, familial hyperlipidemias with high LDL-C and/or TG levels have been
228 reported to be highly polygenic.^{9, 10, 13, 30} The pleiotropic effects of diverse genes, in contrast with the
229 single affected pathway in monogenic FH, may partly explain why we did not observe increased CAD
230 risk due to familiarity in our study.

231 Second, to more deeply characterize potential differences between familial and population-
232 ascertained hyperlipidemias, we performed precise phenotyping of circulating lipid species known to
233 be associated with ASCVD risk.²⁰⁻²² We first characterized the lipid profiles associated with high LDL-
234 C and TG levels. Many of the associations were not specific to LDL-C but rather due to combined
235 dyslipidemia. LDL particles are generated in circulation as downstream metabolic products from the
236 TG-enriched lipoproteins (liver-derived VLDL particles and their post-lipolytic remnants) by the action
237 of two lipases, lipoprotein lipase and hepatic lipase.^{31,32} Thus, a proportion of the core lipids—especially
238 cholesterol esters—and the particle surface phospholipids are retained within the generated LDL
239 particles. The actions of the cholesteryl ester transfer protein and phospholipid transfer protein further
240 modulate the constituents of TG-rich and LDL particles.³³ Percentual lipid compositions have been
241 reported for different lipoprotein classes, but they do not directly reflect variation in plasma LDL-C or
242 TG concentrations. For example, PCs have been estimated to constitute 12 % of all lipids in LDL
243 particles vs. 3–9% in TG-rich lipoproteins.³⁴ However, in our study, PCs were overall more strongly
244 associated with TG levels than with LDL-C levels. Nevertheless, LDL-C remained positively associated
245 with a range of species including CEs, ceramides, SMs, PCs and PCOs. Among the strongly increased
246 species, CE(14:0), CE(16:0), CE(16:1), CE(18:0), SM(34:1;2), SM(34:2;2), SM(42:2;2), Cer(42:1;2),
247 and Cer(42:2;2) have previously been associated with the risk of future ASCVD incidence or events.^{20,}

248 ²¹

249 Elevated TG levels were associated with either increases or decreases in the levels of lipid
250 species across most classes. Importantly, most of these associations appeared to be independent of LDL-
251 C levels. Among the lipid species that were strongly correlated with high TGs even after correction for
252 LDL-C levels were several species which have previously been associated with risk of ASCVD
253 incidence or events.²⁰⁻²² These include the species CE(14:0), CE(16:0), CE(16:1), CE(18:0), TG(50:1),
254 TG(50:2), TG(50:3), TG(52:2), TG(52:3), TG(52:5), TG(56:5), TG(56:6), Cer(42:1;2) and Cer(42:2;2).
255 Further, high TGs were associated with increased saturation of fatty acids in the TG, DG, CE and LPC
256 classes. Such differences in the relative concentrations of circulating fatty acids can be partly related to
257 dietary intake, but are also influenced by endogenous metabolism.³⁵

258 Overall, a larger proportion of the lipid species previously linked with increased ASCVD risk
259 were strongly associated with elevated TGs rather than with elevated LDL-C. This interesting
260 observation suggests that the levels of these lipid biomarkers are more closely linked with circulating
261 TG-rich lipoprotein metabolism than with low-density lipoproteins.

262 Third, several lipid species, such as specific cholesteryl esters, ceramides, and
263 phosphatidylcholine-ethers, remained independently associated with both elevated LDL-C and TGs.
264 Among these species, the ceramides Cer(42:1;2) (presumably Cer(d18:1/24:0)) and Cer(42:2;2)
265 (presumably Cer(d18:1/24:1)), the sterol esters CE(16:1) and CE(18:0), and the sphingomyelin
266 SM(34:1;2) may have added value in the prediction of ASCVD incidence or event risk over traditional
267 lipid measurements.²⁰⁻²² Plasma ceramides have been reported to be independent predictors of
268 cardiovascular events in addition to LDL-C in the population and in CAD patients.^{22, 36, 37} Both LDL-C
269 and TGs remained independently associated with all four ceramides quantified in our study, and LDL-
270 C was additionally associated with increased saturation degree of ceramides. Unlike most CE species,
271 CE(16:1) was more strongly associated with the concentration of TGs than with LDL-C concentration
272 in our study. SM(34:1;2) was the only sphingomyelin species which was negatively associated with TGs
273 – and this association became evident only after adjusting for LDL-C levels. Additionally, some species,
274 such as Cer(42:1;2) and TG(56:6), which were positively associated with hyperlipidemias in our sample,
275 have previously been reported to be associated with decreased risk of ASCVD events.^{20, 22} These co-

276 associations and discordances between reported associations might explain why some lipid species can
277 improve risk prediction. Consequently, there is an urgent need for a better understanding of the potential
278 underlying signaling and metabolic pathways.

279 Finally, the lipidomic profiles associated with high LDL-C or TG levels were comparable
280 between familial and population-ascertained hyperlipidemias. We observed no differences in either the
281 levels of individual lipid species or the saturation of fatty acids within lipid classes. Our results support
282 the hypothesis that familial and population-ascertained hyperlipidemias have similar, overlapping, and
283 heterogeneous pathophysiology. Our results are also reassuring for the quality of lipidomics studies
284 which combine familial and population-based hyperlipidemic samples to increase statistical power.

285 Although we present the most comprehensive characterization of coronary artery disease risk
286 and circulating lipid species in common familial hyperlipidemias to date, our study has some limitations.
287 The confidence intervals for CAD risk were relatively large in the family samples. The field of
288 lipidomics is still relatively young, and concerns have been raised regarding the replicability of
289 individual lipidomics platforms. The platform used here overcomes these problems by profiting from
290 the advantages of direct infusion mass spectrometry for high throughput screening studies. We also
291 observed similar lipidomic profiles associated with hyperlipidemia in two independent cohorts,
292 supporting the replicability of the platform. Further, the lipid species included in our analyses are
293 strongly associated with known genetic lipid loci (Supplemental Figures V–VIII). We profiled the whole
294 circulating lipidome at once to capture variation across various lipoprotein classes and size distributions.
295 However, this limits our ability to follow up on observed differences at the level of the individual
296 lipoproteins. We excluded poorly captured lipid species from the analyses; future advances in lipidomics
297 technology might enable their detection. It is unclear how well our results can be generalized to other
298 populations than Finns. Some of the individuals surveyed in population cohorts might in fact have
299 familial hyperlipidemia; we could not fully rule out such cases. As this was a cross-sectional study,
300 single LDL-C and TG measurements were used to establish hypercholesterolemia status. However,
301 especially TGs are known to exhibit diurnal variability—especially in relation to dietary episodes with
302 chylomicron entry—as well as variation over time periods of months to years.³⁸⁻⁴⁰ In this study, the

303 samples from the hyperlipidemic families were fasting and the samples from the population cohort were
304 semi-fasting. In this light, the similarity of lipidomic profiles between familial and population-
305 ascertained hypertriglyceridemias becomes even more striking. Moreover, the most recent consensus
306 statement from the European Atherosclerosis Society and European Federation of Clinical Chemistry
307 and Laboratory Medicine recommends routine use of non-fasting blood samples for assessment of
308 plasma lipid profiles.⁴¹

309 In conclusion, we observed distinct lipidomic profiles associated with high levels of LDL-C and
310 TGs, but similar CAD risk and lipidomic profiles between common familial and population-ascertained
311 hyperlipidemias, providing further evidence of similar and overlapping underlying mechanisms. The
312 high degree of similarity between these familial dyslipidemic individuals and random population
313 samples in their genomic and lipidomic profiles, and in elevated CAD risk, does not support different
314 screening and treatment for familial and sporadic cases. Additional work is needed to confirm the
315 validity of this hypothesis in clinical settings.

316 **Funding**

317 This work was supported by National Institutes of Health [grant number HL113315 to S.R., M.-R.T.,
318 N.B.F, and A.P.]; Finnish Foundation for Cardiovascular Research [to S.R., V.S., M.-R.T., M.J., and
319 A.P.]; Academy of Finland Center of Excellence in Complex Disease Genetics [grant number 213506
320 and 129680 to S.R. and A.P.]; Academy of Finland [grant number 251217 and 285380 to S.R. and grant
321 number 286500 to A.P.]; Jane and Aatos Erkko Foundation [to M.J.]; Sigrid Jusélius Foundation [to
322 S.R., A.P., and M.-R.T.]; Biocentrum Helsinki [to S.R.]; Horizon 2020 Research and Innovation
323 Programme [grant number 692145 to S.R.]; EU-project RESOLVE (EU 7th Framework Program) [grant
324 number 305707 to M.-R.T.]; HiLIFE Fellowship [to S.R.]; Helsinki University Central Hospital
325 Research Funds [to M.-R.T.]; Magnus Ehrnrooth Foundation [to M.J.]; Leducq Foundation [to M.-R.T.];
326 Ida Montin Foundation [to P.R.]; MD-PhD Programme of the Faculty of Medicine, University of
327 Helsinki [to J.T.R.]; Doctoral Programme in Population Health, University of Helsinki [to J.T.R. and

328 P.R.]; Finnish Medical Foundation [to J.T.R.]; Emil Aaltonen Foundation [to J.T.R. and P.R.];
329 Biomedicum Helsinki Foundation [to J.T.R.]; Paulo Foundation [to J.T.R.]; Idman Foundation [to
330 J.T.R.]; and Veritas Foundation [to J.T.R.]. The funders had no role in study design, data collection and
331 analysis, decision to publish, or preparation of the manuscript.

332 **Acknowledgements**

333 We would like to thank Sari Kivikko, Huei-Yi Shen, and Ulla Tuomainen for management assistance.
334 The FINRISK data used for the research were obtained from THL Biobank. We thank all study
335 participants for their generous participation in the FINRISK and EUFAM studies.

336 **Conflicts of Interest**

337 K.S., C.K., and M.J.G. have paid employment at Lipotype GmbH. This does not alter the authors'
338 adherence to all policies on sharing data and materials.

339

340

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493

494

495

496 **Tables**

497 **Table 1. Coronary artery disease risk in familial and population-ascertained hyperlipidemias.**

Hyperlipidemia type	Parameter	Hyperlipidemic families	Population cohort	Meta-analysis
High LDL-C	<i>n</i> (hyperlipidemic / non-hyperlipidemic)	625 (136/489)	19,644 (2,175/17,469)	20,269 (2,311/17,958)
	Number of incident CAD diagnoses (hyperlipidemic / non-hyperlipidemic)	45 (16/29)	904 (176/728)	949 (192/757)
	CAD HR (95% CI)	1.71 (0.94-3.10)	1.74 (1.48-2.05)	1.74 (1.48-2.04)
High TGs	<i>n</i> (hyperlipidemic / non-hyperlipidemic)	371 (72/299)	19,644 (1,405/18,239)	20,015 (1,477/18,538)
	Number of incident CAD diagnoses (hyperlipidemic / non-hyperlipidemic)	21 (3/18)	904 (74/830)	925 (77/848)
	CAD HR (95% CI)	1.35 (0.52-3.51)	1.38 (1.08-1.75)	1.38 (1.09-1.74)

498

499 *Clinical incident CAD event endpoints were defined as either myocardial infarction or coronary revascularization.*

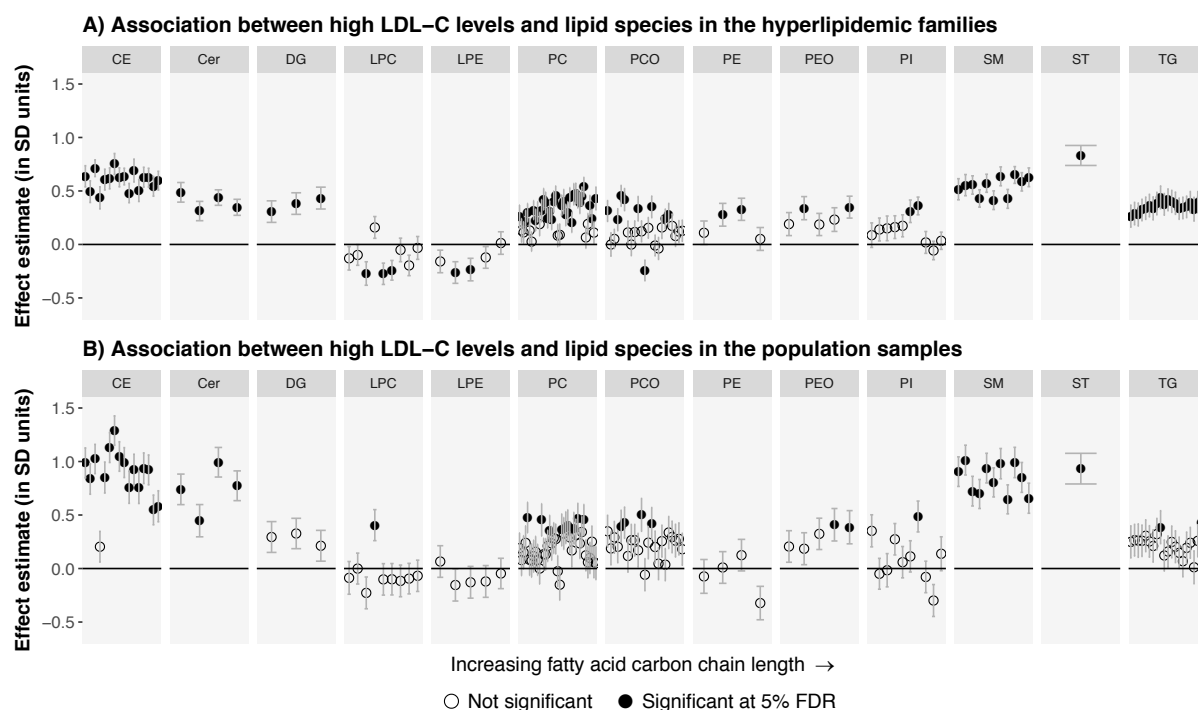
500 *Cox proportional hazards models, stratified by sex and clustered by family were used to estimate hazard ratios*

501 *(HR) for incident CAD events using age as the time scale and excluding individuals with prevalent CAD.*

502

503 **Figures and Figure Legends**

504 **Figure 1. Associations between high LDL-C status and the levels of 151 lipid species.**



505

506 Panel A), individuals affected by high LDL-C levels ($n=105$) were compared with their unaffected

507 relatives ($n=358$) in the 53 “high LDL-C” families. Panel B), individuals affected by high LDL-C ($n=56$)

508 were compared with other individuals ($n=841$) in the FINRISK population cohort. The association of

509 high LDL-C status with the lipid species was estimated using linear mixed models with age, age², and

510 sex as the other fixed effect covariates. Statistical significance was evaluated using the Benjamini-

511 Hochberg method at a 5% false discovery rate (FDR). The ordering of the lipid species within each class

512 is the same as in Supplemental Table III. *Cer* = ceramide, *DG* = diacylglyceride, *LDL-C* = low-density

513 *lipoprotein cholesterol*, *LPA* = lysophosphatic acid, *LPC* = lysophosphatidylcholine, *LPE* =

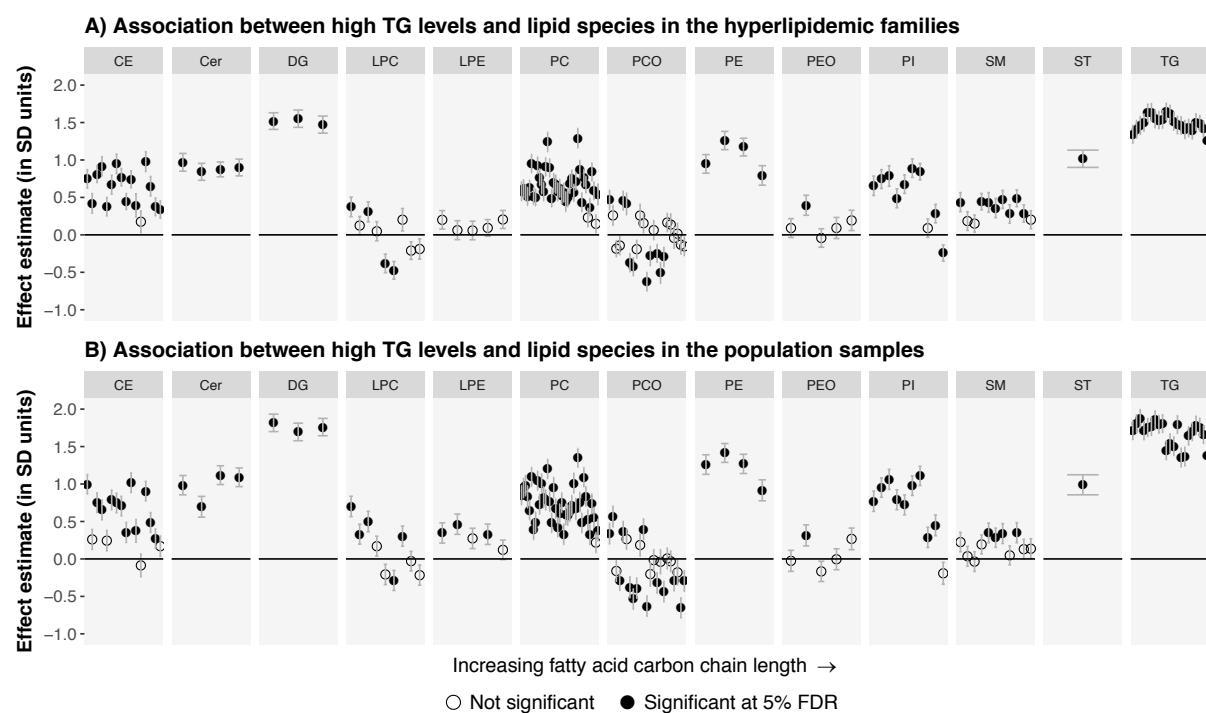
514 *lysophosphatidylethanolamine*, *PC* = phosphatidylcholine, *PCO* = phosphatidylcholine-ether, *PE* =

515 *phosphatidylethanolamine*, *PEO* = phosphatidylethanolamine-ether, *PI* = phosphatidylinositol, *CE* =

516 *cholesteryl ester*; *SM* = sphingomyelin, *ST* = sterol, *TG* = triacylglyceride.

517

518 **Figure 2. Associations between high TG status and the levels of 151 lipid species.**

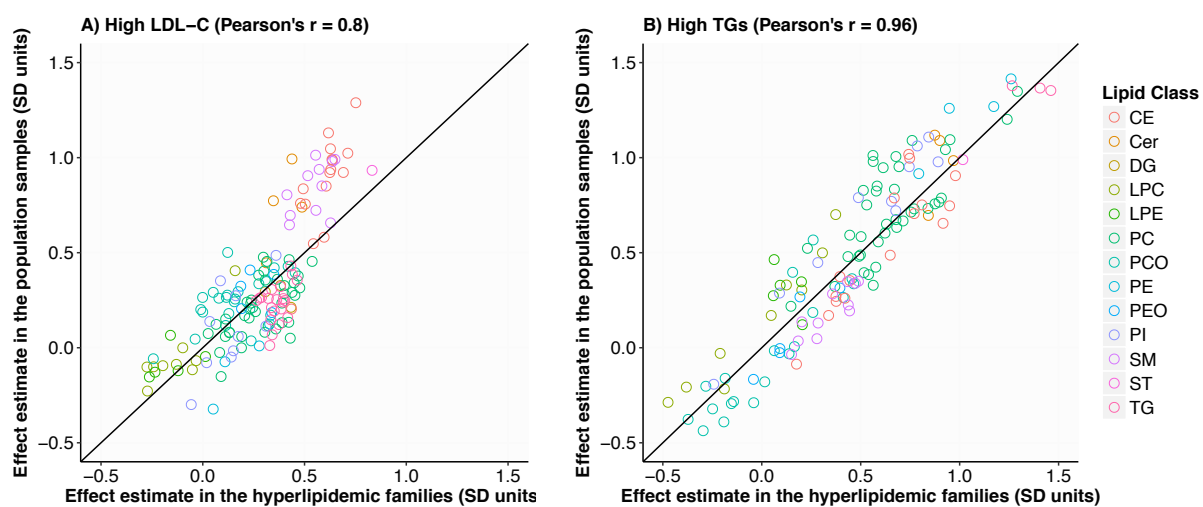


519

520 Panel A), individuals affected by high TGs ($n=64$) were compared with their unaffected relatives
521 ($n=223$) in 39 “high TG” families. Panel B), individuals affected by high TGs ($n=65$) were compared
522 with other individuals ($n=832$) in the FINRISK population cohort. The association analyses were
523 performed similarly to Figure 1. Abbreviations are displayed in Figure 1 legend.

524

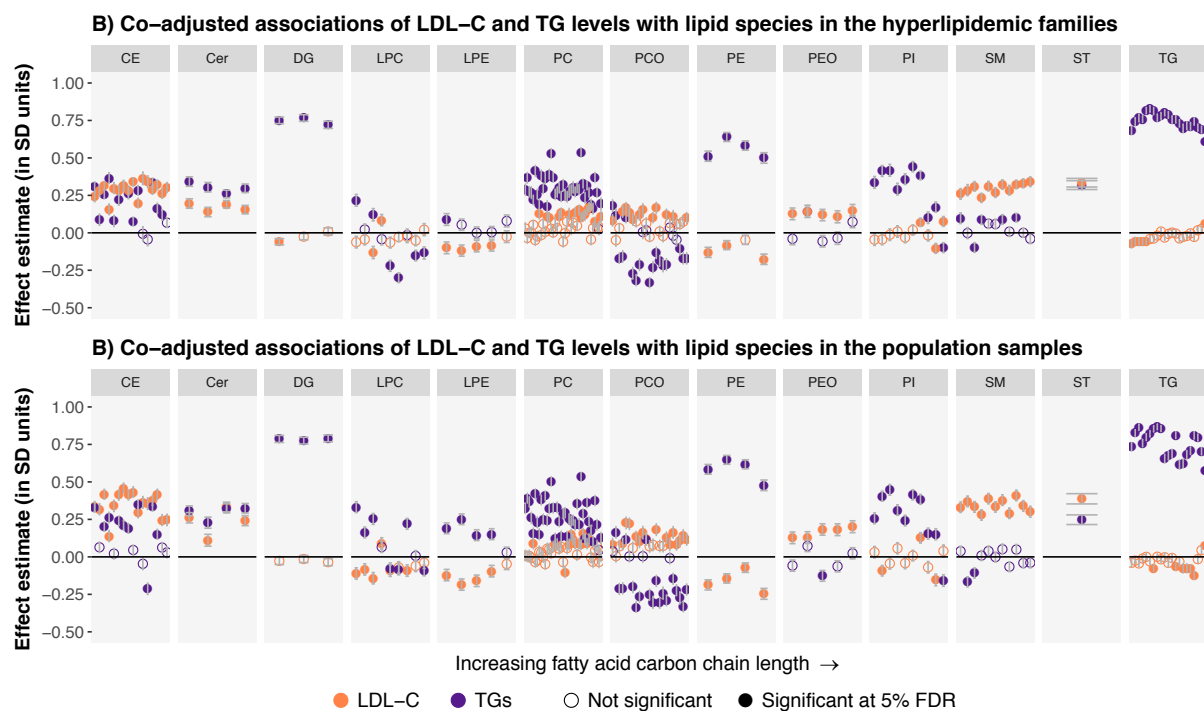
525 **Figure 3. Correlation of effect estimates for hyperlipidemia status between the hyperlipidemic**
526 **families and the population samples.**



528 The correlation between the effect estimates observed in the family and population cohorts is presented
529 in A) for high LDL-C (effect estimates presented in Figure 1) and B) for high TGs (effect estimates
530 presented in Figure 2). Abbreviations are displayed in Figure 1 legend.

531

532 **Figure 4. Independent (co-adjusted) associations of LDL-C and TGs with 151 lipid species.**



533

534 Effect estimates for LDL-C and TGs were derived from linear mixed models with the lipid species as
535 outcomes, and LDL-C, $\log(TGs)$, age, age^2 , and sex as fixed effect covariates. The effect estimates were
536 derived separately in the hyperlipidemic families (Panel A, $n = 550$ individuals) and the FINRISK
537 population cohort (Panel B, $n = 897$ individuals). Effect estimates are presented for LDL-C in orange
538 and TGs in purple. Statistical significance was evaluated using the Benjamini-Hochberg method at a 5%
539 false discovery rate (FDR). The ordering of the lipid species within each class is the same as in
540 Supplemental Table III. Abbreviations are displayed in Figure 1 legend.