Fine mapping of GWAS signals to identify genetic markers of the plasma triglyceride response to an omega-3 fatty acid supplementation

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Short running head: Genetic markers of the triglyceride response

Abbreviations:

BMI: Body mass index; DHA: docosahexaenoic acid; EPA: eicosapentaenoic acid; FA: fatty acid; *FAS* study: *Fatty Acid Sensor* study; GRS: genetic risk score; GWAS: genome-wide association study; n-3: omega-3; SNPs: single nucleotide polymorphisms; TG: triglyceride.

Clinical Trial Registry number and website (FAS study): <u>http://www.clinicaltrials.gov</u> (NCT01343342)

1 ABSTRACT

Background: Using a genome-wide association study (GWAS) approach, our group
previously computed a genetic risk score (GRS) from single nucleotide polymorphisms
(SNPs) of ten loci which affect the plasma triglyceride (TG) response to an omega-3 (n-3)
fatty acid (FA) supplementation.

Objective: The objective was to compute a novel and more refined GRS using fine
mapping to include a large number of genetic variants.

Design: A total of 208 participants of the *Fatty Acid Sensor (FAS)* study received 5g of fish oil per day, containing 1.9–2.2g of eicosapentanoic acid and 1.1g of docosahexanoic acid, for six weeks. Plasma TG levels were measured before and after supplementation. Dense genotyping and genotype imputation were employed to refine mapping around GWAS hits. A GRS was computed by summing the number of at-risk alleles of tagging SNPs. Analyses were replicated in samples of the *FINGEN* study.

Results: A total of 31 tagging SNPs associated with the TG response were used for GRS calculation in the *FAS* study. In a general linear model adjusted for age, sex and body mass index, the GRS explained 49.73% of TG response variance (p < 0.0001). Nonresponders to the n-3 FA supplementation had a higher GRS than responders. In the *FINGEN* replication study, the GRS explained 3.67% of TG response variance (p = 0.0006).

20 **Conclusion:** Fine mapping proved to be effective to refine the previous GRS. Carrying 21 increasing numbers of at-risk alleles of 31 SNPs confers a higher risk of being non-22 responsive to n-3 FA. The genetic profile therefore appears to be an important determinant of the plasma TG response to an n-3 FA supplementation and could be used to target
those most likely to gain clinical benefit.

- 25 Keywords: genetic risk score, plasma lipid levels, omega-3 fatty acids, genome-wide
- 26 association study, nutrigenetics, gene-diet interactions.

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28 BACKGROUND

The hypotriglyceridemic effect of marine omega-3 (n-3) fatty acids (FA) has been 29 consistently described (1). At a population level, a daily intake of 4g of n-3 FA in the form 30 of eicosapentaenoic acid (EPA) and docosahexaenoic acid (DHA) can lower triglyceride 31 32 (TG) levels by up to 30% (1). However, it remains unclear whether n-3 FA consumption actually translates into a decrease of cardiovascular events and mortality risk, for 33 discrepancies have been reported in the literature (2-5). A possible explanation for these 34 inconsistent results is that not all individuals equally benefit from n-3 FA consumption. 35 Accordingly, a large inter-individual heterogeneity in the plasma TG response to an n-3 36 FA supplementation has been observed by many research groups. In the FINGEN study 37 40% of participants who underwent an 8-week n-3 FA supplementation did not show a 38 decrease in TG levels (6). Likewise, 29% of participants of the Fatty Acid Sensor (FAS) 39 study, conducted by our research group, did not have decreased TG levels after a 6-week 40 supplementation of 3g of n-3 FA a day, comprising 1.9-2.2 g of EPA and 1.1 g DHA (7). 41

The etiology of the variability in the hypotriglyceridaemic response is likely to be multi-42 factorial, with genetic factors partly accounting for the inter-individual variability of the TG 43 44 response to an n-3 FA supplementation (8). Our group previously conducted a genomewide association study (GWAS) on participants of the FAS study to identify potential 45 variants associated with the plasma TG response to n-3 FA supplementation and 46 47 identified 13 loci located in six genes, namely IQCJ-SCHIP1, NXPH1, PHF17, MYB, NELL1 and SLIT2 (9). A genetic risk score (GRS) was computed from ten GWAS hits and 48 49 this explained 21.53% of the TG response (9). We recently reported several gene-diet 50 interactions modulating TG levels following the n-3 FA supplementation after increasing the density of markers around GWAS hits by dense genotyping (10). These results 51

demonstrated the importance of pursuing mapping refinement around GWAS-associated loci to identify actual causative single nucleotide polymorphisms (SNPs). It is therefore very likely that the current GRS on *FAS* participants could be improved by using fine mapping to add variants, which would bring more power and accuracy.

The objective of the present study was to generate a more refined and improved GRS using fine mapping of GWAS hits regions to add SNPs in order to more accurately predict the individual TG response to an n-3 FA supplementation.

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62 MATERIALS AND METHODS

63 Study population

A total of 254 healthy Canadian Caucasian subjects living in the Quebec City metropolitan 64 65 area were recruited into the FAS study from September 2009 to December 2011 using announcements in local newspapers, as well as electronic messages sent to students and 66 employees of Université Laval (Quebec). Inclusion criteria were as follows: aged between 67 18 and 50 years; a body mass index (BMI) between 25 and 40 kg/m²; non-smokers; free 68 from thyroid or metabolic disorders requiring a treatment such as diabetes, hypertension, 69 dyslipidemia or cardiovascular disease. Subjects were not eligible if they had taken n-3 70 FA supplements for a minimum of six months prior to the intervention. A total of 210 71 participants completed the intervention. However, two participants did not have plasma 72 73 TG levels data available for further analyses and were therefore excluded, yielding a final sample of 208 participants. Subjects were subsequently separated into two subgroups: 1-74 responders to the n-3 FA supplementation; and 2- non-responders. Responders were 75 76 defined as participants whose plasma TG levels decreased after the n-3 FA supplementation, whereas non-responders were participants whose TG levels remained 77 stable or increased through the n-3 FA supplementation. 78

79 Study design and diets

The study design and diets have been previously reported (9, 11). Flow chart of participants and intervention is presented as online supporting material (**Supplemental Figure 1**). First, participants followed a run-in period of two weeks, where they were given dietary instructions by a trained registered dietitian to achieve the recommendations from the *Canada's Food Guide*. The purpose of these recommendations was to ensure that

participants had a constant n-3 FA dietary intake and maintained a stable body weight 85 throughout the study period. More specifically, they were asked not to consume more than 86 150 g of fish or seafood per week, to avoid food enriched with n-3 FA, to limit alcohol 87 consumption to a maximum of two regular drinks per week and to avoid any dietary 88 supplement (n-3 FA supplements, vitamins or natural products) during the intervention. 89 Thereafter, participants were asked to consume the n-3 FA capsules daily for six weeks. 90 (Ocean Nutrition, Nova Scotia, Canada). The five capsules per day provided a total of 3 g 91 of n-3 FA a day, including 1.9 – 2.2 g of EPA and 1.1 g of DHA. Participants were asked 92 to report any deviation from the protocol. They were also asked to record any experienced 93 side effects, alcohol intake and fish consumption. 94

95 Laboratory methods

96 Plasma lipids

97 Methods used to measure blood lipids have already been published (7). Briefly, blood 98 samples were collected after a 12h overnight fast and 48h alcohol abstinence at the 99 beginning and end of the intervention. Plasma TG concentrations were assessed by 100 enzymatic assays (12, 13).

101 Gene mapping

Two methods were used for fine mapping: dense genotyping and genotype imputation. For genotyping, the whole procedure was previously described (10). Briefly, SNPs were identified using the International HapMap Project SNP database, based on the National Center for Biotechnology information (NCBI) B36 assembly Data Release 28, phase II + III, built 126. Tagging SNPs were selected using the Tagger procedure in Haploview v4.2

according to their minor allele frequency (MAF >5%) and pairwise tagging ($r^2 \ge 0.80$). A 107 108 total of 16 SNPs in IQCJ, 34 in NXPH1, 8 in PHF17 and 9 in MYB were chosen to cover all common variations at these chromosomal regions. To the 16 SNPs in IQCJ, 23 109 genotyped SNPs in SCHIP1 were added in order to cover the full IQCJ-SCHIP1 gene. All 110 SNPs were in Hardy-Weinberg Equilibrium. The GenElute Gel Extraction Kit (Sigma-111 Aldrich Co., St. Louis, MO) was used to extract genomic DNA (gDNA) from the blood 112 samples. Genotyping was conducted by polymerase chain reaction (PCR) using TagMan 113 technology (Life Technologies, Carlsbad, CA) in 210 subjects. 114

For genotype imputation, the 1000 Genomes project data (release 1000G Phase I v3, 115 updated 26 Aug 2012) was used as a reference set for the imputation of genotypes 116 (genotyped from Illumina BeadChip) of previously identified GWAS loci (9). A total of 1684 117 markers in IQCJ-SCHIP1, 1684 in NXPH1, 885 in PHF17 and 777 in MYB, that were 118 originally used to conduct the GWAS in 141 participants, were used to infer other 119 genotypes using algorithms implemented in IMPUTE2 (14). The imputation cut-off was 120 0.90, with 99.0% of success rate. A total of 52770 informative SNPs in IQCJ-SCHIP1. 121 50218 in NXPH1, 30140 in PHF17, 29725 in MYB, 61560 in NELL1 and 61736 in SLIT2 122 were obtained from genotype imputation, including initial markers. 123

Imputed SNPs were then submitted to quality control tests, where only polymorphic SNPs with a genotype call rate >95% and MAF \geq 1% were conserved. Quality control tests left 5205 SNPs in *IQCJ-SCHIP1*, 6040 in *NXPH1*, 3028 in *PHF17*, 2616 in *MYB*, 7846 in *NELL1* and 7124 in *SLIT2* available for SNP analysis.

128 SNP analysis and selection

Allele frequency between responders and non-responders was calculated and compared using PLINK. Odds ratio reporting the ratio between the proportion of non-responders carrying the minor allele of a SNP and the proportion of responders carrying the minor allele of the same SNP was calculated. Odds ratio *P* values were calculated using a Chisquare test.

Prior to statistical analyses, tagging SNPs were selected using PLINK from genotyping and imputation data. For inclusion criteria, the r^2 was set at ≥ 0.80 and the *P* value was 0.05. A final sample of 88 independent tagging SNPs in *IQCJ-SCHIP1*, 88 in *NXPH1*, 56 in *PHF17*, 97 in *MYB*, 58 in *NELL1* and 118 in *SLIT2* were kept for statistical analyses.

138 *Replication study*

Analyses were replicated in the European FINGEN study, a trial conducted at the 139 Universities of Glasgow, Newcastle, Reading and Southampton in the United Kingdom 140 from 2003 to 2005. Over 95% of participants were British Caucasians. The study design 141 has been previously published (6, 15). Briefly, it was a double-blind, placebo-controlled, 142 dose-response crossover study in which participants received either a placebo, 0.7 g of 143 144 EPA and DHA a day or 1.8 g of EPA and DHA a day for eight weeks with 12-week washouts in between. Responsiveness to n-3 FA supplementation was defined in the 145 same way as outlined above for the FAS participants, with only the response to the 146 147 1.8g/day dose used in the current analysis. TG data was available for 310 of the 312 148 subjects. Genotyping was conducted by LCG genomics (16).

149 Statistical analyses

Statistical analyses were conducted using SAS software v9.4. Normal distribution was evaluated with the box-plot, skewness and kurtosis ranges. Abnormally distributed variables were log_{10} -transformed. Statistical significance was set at *P* < 0.05. An unpaired t-test was used to assess differences between responders and non-responders prior to the supplementation.

A GRS was calculated for each participant from the sum of risk alleles of tagging SNPs in an additive way. To select the most relevant SNPs to include in the GRS calculation, stepwise bidirectional regressions adjusted for age, sex and BMI were conducted in each gene separately (REG procedure in SAS) to assess the contribution of SNPs to the TG variation (Δ TG). Significant SNPs (p<0.05) were kept for the GRS calculation.

A *t*-test was performed to compare mean GRS scores between responders and nonresponders to the n-3 FA supplementation. A general linear model (GLM procedure in SAS) adjusted for age, sex and BMI was used to assess the effect of the GRS on the plasma TG response to the n-3 FA supplementation. Sensitivity and specificity of the GRS were calculated by measuring the area under the receiver operating characteristic (ROC) curve using the logistic procedure in SAS with adjustments for age, sex, BMI and baseline TG levels.

167 **RESULTS**

Characteristics of participants and FA profiles were previously reported (9). Table 1 168 presents a summary of baseline and post-supplementation characteristics of responders 169 and non-responders to the n-3 FA supplementation. Participants were overweight, with 170 mean (SD) baseline BMIs of 28.9 (3.6) kg/m² and 27.8 (3.9) kg/m² in the responder and 171 non-responder groups, respectively. Responders had higher TG levels at baseline 172 compared to non-responders (p < 0.0001), and their TG levels significantly decreased by 173 on average 0.50 mmol/l through supplementation (p < 0.0001), whereas non-responders 174 had a mean 0.17 mmol/l increase (p < 0.0001). 175

Figure 1 shows *p* value for differences in allele frequency between responders and nonresponders in GWAS-associated genes after mapping refinement by genotype imputation (A: *IQCJ-SCHIP1*, B: *SLIT2*, C: *PHF17*, D: *MYB*, E: *NXPH1*, F: *NELL1*). A total of 62 markers passed the significance threshold used in the GWAS ($p = 10^{-5}$), counting 12 in *IQCJ-SCHIP1*, one in *NXPH1*, 22 in *PHF17*, six in *MYB*, one in *NELL1* and 20 in *SLIT2*.

Six stepwise bidirectional models (one for each gene) adjusted for age, sex and BMI were 181 182 conducted to select SNPs to include in the GRS calculation. According to the stepwise models, 31 SNPs, namely two in IQCJ-SCHIP1, 10 in NXPH1, three in PHF17, four in 183 *MYB*, four in *NELL1* and eight in *SLIT2*, were associated with the TG response (**Table 2**) 184 185 which were used in the GRS calculation. Figure 2 presents the GRS distribution in the study population. The higher the GRS score is, the more a subject carries at-risk alleles. 186 Responders had lower GRS score (1.26 ± 2.34) in comparison to non-responders (6.32 ± 187 188 2.21) (p < 0.0001). In a general linear model adjusted for age, sex and BMI, the GRS was significantly associated with the TG response (p < 0.0001). The GRS accounted for 49.73% of TG change following the n-3 FA supplementation (p < 0.0001).

191 Sensitivity and specificity of the genetic risk model were assessed with the area under the ROC curve. The area under the curve was 0.9366 [95% confidence interval (CI): 0.8976. 192 0.9756] for the GRS solely, and 0.7537 [95% CI: 0.6721, 0.8353] for the sum of other 193 determinants (general model), including BMI, sex, age and baseline TG levels. The 194 addition of the GRS to the general model significantly increased the predictive power 195 (p<0.0001), for an area under the curve of 0.9455 [95% CI: 0.9084, 0.9826] (Figure 3). 196 After cross-validation, the area under the curve was 0.9187 for the GRS alone and 0.9280 197 for the full adjusted model (data not shown). 198

As to the FINGEN replication study, there were 122 non-responders (39.7%) and 188 199 200 responders (60.3%), according to the same definition as mentioned above. Genotyping in FINGEN participants was conducted on the 31 SNPs of the GRS. Among these 31 SNPs, 201 eight were either monomorphic in the FINGEN cohort (rs61569932, rs1216346, 202 203 rs79624996, rs10009535 and rs76015249) or failed genotyping (rs6966968, rs78943417 and rs184945470), leaving 23 SNPs for GRS calculation. In the general linear model 204 adjusted for age, sex and BMI, this GRS was also significantly associated with the TG 205 response (p = 0.0006) and accounted for 3.67% of the TG change following the n-3 FA 206 207 supplementation (p = 0.0006). Regarding sensitivity and specificity, the areas under the curve were 0.6417 [95% CI: 0.5795, 0.7039] for the GRS solely, 0.7109 [95% CI): 0.6530, 208 0.7688] for the general model and 0.7553 [95% CI: 0.7010, 0.8095] for the general model 209 + GRS (Figure 4). 210

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212 **DISCUSSION**

The present study follows up a GWAS of the plasma TG response to an n-3 FA 213 supplementation in which a GRS was constructed (9). We used dense genotyping, as well 214 as genotype imputation to further increase the density of markers identified in the GWAS. 215 Association studies were performed and a new GRS was computed from imputed and 216 genotyped tagging SNPs to better predict the plasma TG response to an n-3 FA 217 supplementation. The previous GRS was computed using a total of ten GWAS hits (9). 218 219 Using genotype imputation and dense genotyping data, we were able to construct a more refined GRS using 31 SNPs. As illustrated in Figure 2, we now have a clear disparity 220 between responders and non-responders regarding their number of carried at-risk alleles. 221 222 Non-responders have much higher GRS levels than responders. The 31-SNPs GRS explains a much larger proportion of the TG variance (49.73%) than the 10-SNPs GRS 223 224 (21.53%) and its predictive capacity for classifying individuals into responders and nonresponders categories is also highly accurate. However, the GRS poorly explained TG 225 variance in the FINGEN population (3.67%) compared to FAS. This is consistent with 226 previous results of the 10-SNPs GRS, where only 2% of the TG variation was explained 227 by the GRS in the FINGEN population (nonsignificant) compared to 21.53% in the FAS 228 population (p = 0.0002) (9). This important divergence could be explained by the eight 229 missing SNPs in the GRS in the replication study. Also, differences in allele frequency 230 between the two populations might be a contributing factor. Accordingly, several SNPs 231 (rs62270407, rs10009109, rs76015249, rs61569932 and rs293180) showed significant 232 233 differences in allele frequency between the FAS and FINGEN populations. These differences can be explained by their ancestry background. FAS study participants are 234

French-Canadians of european descent, a more homogeneous population with a founder effect while over 95% of *FINGEN* study participants were British Caucasians (17). Moreover, the proportion of non-responders among *FAS* participants is 29% vs almost 40% among *FINGEN* participants.

Despite the clear influence of SNPs on the responsiveness of TG levels to n-3 FA 239 supplementation, the exact mechanisms by which the six genes in GWAS-associated loci 240 contribute to TG variation still remain unclear. We previously demonstrated that SNPs in 241 GWAS-associated loci may exert their effect on TG levels by influencing gene expression 242 via modulation of DNA methylation (18). However, most of these genes are poorly 243 connected to lipid metabolism as detailed in our previous paper (10). This is especially 244 245 the case for IQCJ-SCHIP1. Nevertheless, even though IQCJ-SCHIP1 has not been explicitly linked to lipid metabolism, we hypothesize that its action on the TG response 246 could be mediated through calcitonin action. 247

Accordingly, Martin PM et al. demonstrated that, in absence of calcium, IQCJ binds 248 249 calmodulin via its motif IQ, a sequence motif not shared with the SCHIP1 segment (19, 20). Calmodulin is a messenger protein that normally binds to calcium to modulate its 250 action (21). However, other proteins, like IQ motif-containing proteins, also interact with 251 calmodulin in a calcium-independent way, more specifically when calcium levels are low 252 253 (21). Calmodulin is implicated in lipid metabolism via its effect on calcitonin (22), a thyroid hormone that regulates calcium and phosphate in the blood (23). In a previous study, it 254 255 has been reported that the administration of calcitonin decreased LDL-cholesterol and TG levels in rats, while the administration of a calmodulin inhibitor prevented this effect, thus 256 suggesting that calmodulin suppressed the action of calcitonin (22). Werner and Low also 257

observed that calcitonin inhibited lipolysis during basal and stimulated lipolysis with 258 259 parathyroid hormone, noradrenaline or dibutyryl cAMP (bucladesine) in rat adipose tissue (24). One of the main mechanism of action of calcitonin lies in the capacity of its receptor, 260 a member of G-protein coupled receptors, to couple with the cyclic adenosine 261 monophosphate (cAMP) signal transduction (25). cAMP is an ubiquitous second 262 messenger implicated in lipid metabolism by activating cAMP-dependent protein kinase, 263 an enzyme that enhances TG hydrolysis in adipocytes (26, 27). cAMP-dependent 264 mechanisms also inhibit lipolysis during refeeding (27). Moreover, dietary compounds, 265 including calcium, caffeine and ethanol, can affect adipocyte lipolysis through modulation 266 of cAMP levels, with increased cAMP levels resulting in stimulated lipolysis (27). 267

268 Another possible explanation is the presence of linkage disequilibrium between tagging SNPs and other SNPs located in transcriptional units other than GWAS-associated loci. 269 As shown in **Table 3**, the majority of the 31 tagging SNPs are not located within GWAS-270 associated loci, but are rather intergenic, and some GWAS-associated loci are located 271 close to genes of interest. For instance, NXPH1 is located next to ICA1, a gene involved 272 in the pathogenesis of type 1 diabetes (28). A SNP of ICA1, along with several SNPs of 273 *NXPH1*, were identified in a GWAS of childhood obesity in an hispanic population (29). 274 These results not only highlight the importance of refining GWAS signals to properly 275 identify the most causative SNPs, but also reinforce the hypothesis that genetic profile is 276 a significant determinant of the metabolic response to an n-3 FA supplementation, and 277 these observations could probably extend to other dietary interventions. It is therefore 278 279 crucial to stop overlooking genetic factors for the assessment of responsiveness to such interventions. For instance, a recently published meta-analysis of randomized clinical 280

trials on the association between n-3 FA supplement use and cardiovascular disease 281 282 concluded that n-3 FA consumption was not associated with a reduction of coronary heart disease or major vascular events (30). Another recent meta-analysis of randomized, 283 double-blind, placebo-controlled trials on the efficacy of n-3 FA consumption for the 284 treatment of hypertriglyceridemia concluded that n-3 FA were ineffective (31). Although n-285 3 FA consumption overall significantly reduced TG levels, authors asserted that evidence 286 regarding the effectiveness of n-3 FA in the management of dislipidemia, especially 287 hypertriglyceridemia, is inconclusive, in part because of the heterogeneity in studies, low 288 methodological quality of studies and small sample sizes (31). As raised by Calder P. in 289 290 a commentary on the meta-analysis, their findings are very consistent with literature and actually clearly confirm that n-3 FA in supplemental form can lower TG levels (32). 291 Moreover, this inconclusiveness regarding the beneficial effects of n-3 FA intake can 292 probably be attributable to the lack of consideration of contributors of the interindividual 293 variability in the metabolic response to n-3 FA including genetic factors. Based on our 294 results, it appears that not all individuals can benefit from the TG lowering effects of n-3 295 FA supplements. Future research should focus on addressing the effects of n-3 FA on 296 responders and non-responders separately by stratifying subjects according to their TG 297 298 response and taking determinants including genetic factors into account.

In conclusion, this study further demonstrates the relevance of refining GWAS hits, here
providing a novel, refined GRS highly predictive of the responsiveness to n-3 FA. Genetic
profile appears to be a major determinant of the TG response to n-3 FA supplementation.
Future studies on n-3 FA and other nutrients should pay more attention to the importance
of genetic factors on the inter-individual variability in lipid responsiveness.

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309 Authors' contributions

BVM conducted genotyping, SNP analysis and wrote the paper; BVM and FG conducted genotype imputation and statistical analysis; IR, SL and MCV designed research; PC was responsible for the medical follow-up; PCC and AMM contributed data from the *FINGEN* study; BVM and MCV have primary responsibility for final content. All authors read and approved the final manuscript. Authors declare no conflict of interest.

315 **Consent**

The *FAS* study was approved by the Université Laval and CHU de Québec ethics committees and was performed in accordance with the principles of the Declaration of Helsinki. All participants provided written, informed consent.

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TABLES

Table 1. Characteristics of subjects pre- and post-supplementation (n=141)

	Responders (n=8	31)		Non-responders (n=60)			
Characteristics	Pre- Post- supplementation		P ¹	Pre- supplementation	Post- supplementation	P ¹	
Sex (men/women)	38/43	-	-	30/30	-	-	
Age (years) ²	31.9 ± 8.8	-	-	31.1 ± 9.0	-	-	
Body mass index (kg/m ²) ²	28.9 ± 3.6	28.9 ± 3.7	0.7	27.8 ± 3.9	28.0 ± 4.0	0.0007	
Triglycerides (mmol/l) ²	1.53 ± 0.74	1.03 ± 0.56	< 0.0001	1.03 ± 0.48	1.20 ± 0.54	< 0.0001	

¹ TTEST procedure (SAS v9.4) was used to assess differences pre- vs post-supplementation in responders and nonresponders

² Mean ± standard deviation

	SNP, rs number	Position, base pairs	Location	Alleles ¹	Minor allele frequency			
Gene					Responders	Non-responders	P ²	OR [95%CI] ³
IQCJ-SCHIP1	rs7639707	159148087	Intron	A /G	0.019	0.075	0.020	4.297 [1.138, 16.231]
	rs62270407	159597626	Intron	C/ T	0.352	0.192	0.003	0.437 [0.25, 0.763]
NXPH1	rs61569932	8299207	Upstream <i>NXPH1</i> , intron of <i>ICA1</i>	G /T	0	0.025	0.045	0
	rs1990554	8344530	Upstream <i>NXPH1</i> , intron of <i>ICA1</i>	A /C	0	0.025	0.043	0
	rs6463808	8476787	Intron	A /G	0.086	0.300	<0.0001	4.531 [2.312, 8.881]
	rs6966968	8840378	Downstream <i>NXPH1</i> , intergenic	A/ G	0.111	0.220	0.013	2.261 [1.174, 4.354]
	rs28473103	8842073	Downstream <i>NXPH1</i> , intergenic	A/G	0.420	0.283	0.018	0.546 [0.33, 0.906]
	rs28673635	8855531	Downstream <i>NXPH1</i> , intergenic	A /G	0.111	0.208	0.025	2.105 [1.089, 4.068]
	rs12702829	9049555	Downstream <i>NXPH1</i> , intergenic	C /T	0.365	0.509	0.018	1.797 [1.105, 2.922]
	rs78943417	9062499	Downstream <i>NXPH1</i> , intergenic	A/ AT	0.144	0.034	0.003	0.213 [0.071, 0.633]
	rs293180	9159909	Downstream <i>NXPH1</i> , intergenic	G/ T	0.063	0.158	0.010	2.784 [1.243, 6.234]
	rs1837523	9201284	Downstream <i>NXPH1</i> , intergenic	C /T	0.317	0.188	0.018	0.499 [0.279, 0.891]
PHF17	rs1216346	129555929	Upstream <i>PHF17</i> , intergenic	C /T	0.231	0.526	<0.0001	3.694 [2.196, 6.211]

Table 2. Characteristics of the 31 SNPs used to compute the genetic risk score (GRS)

	rs114348423	130112033	Downstream <i>PHF17</i> , intergenic	A /G	0.006	0.042	0.041	6.998 [0.807, 60.716]
	rs75007521	130286406	Downstream <i>PHF17</i> , intergenic	G /T	0.049	0	0.014	-
MYB	rs72560788	135200886	Upstream <i>MYB</i> , intergenic	C/ T	0.117	0.050	0.049	0.396 [0.153, 1.024]
	rs72974149	135395122	Upstream <i>MYB</i> , intergenic	A/ G	0.130	0.042	0.012	0.292 [0.107, 0.798]
	rs210962	135503785	Intron	C/T	0.303	0.183	0.023	0.518 [0.292, 0.917]
	rs6933462	135584967	Downstream <i>MYB</i> , intergenic	C /G	0.106	0.200	0.028	2.103 [1.073, 4.122]
NELL1	rs79624996	20211262	Upstream <i>NELL1</i> , intergenic	A /G	0.092	0.181	0.032	2.179 [1.055, 4.498]
	rs1850875	20731343	Intron	C /T	0.340	0.533	0.001	2.223 [1.37, 3.608]
	rs78786240	20735026	Intron	C/ T	0.045	0	0.021	-
	rs117114492	21008313	Intron	G /T	0	0.042	0.009	0
SLIT2	rs184945470	19334808	Upstream <i>SLIT2</i> , intergenic	C/ T	0.069	0.158	0.016	2.548 [1.163, 5.583]
	rs143662727	19634162	Upstream <i>SLIT2</i> , intergenic	A/ G	0.037	0	0.033	-
	rs10009109	19655475	Upstream <i>SLIT2</i> , intergenic	C /T	0.388	0.534	0.015	1.811 [1.119, 2.931]
	rs10009535	19747014	Upstream <i>SLIT2</i> , intergenic	A/ G	0.407	0.558	0.012	1.839 [1.141, 2.964]
	rs61790364	19921757	Upstream <i>SLIT2</i> , intergenic	A /G	0.117	0.258	0.002	2.621 [1.397, 4.919]
	rs73241936	20008049	Upstream <i>SLIT2</i> , intergenic	C /T	0.111	0.200	0.038	2.000 [1.03, 3.883]

rs16869663	20485683	Intron	A/ G	0.043	0.110	0.032	2.741 [1.059, 7.102]
rs76015249	20735742	Downstream <i>SLIT2</i> , intron of <i>KCNIP4</i>	A /G	0	0.025	0.043	0

¹ Minor allele in bold

² P values for differences in allele frequency between responders and non-responders were assessed using a Chi square test in PLINK.

³ Odds ratio reporting the ratio of the proportion of non-responders carrying the minor allele over the proportion of responders

FIGURE TITLES/LEGENDS

Figure 1. Manhattan plot showing *p* values for differences in allele frequency between responders and non-responders to omega-3 fatty acid supplementation in each gene region identified by the initial genome-wide association study (GWAS) (9). SNPs obtained from genotype imputation are shown in red in panels below. Differences in allele frequency were assessed using a Chi square test in PLINK.

Figure 2. Genetic risk score (GRS) distribution in study population (n=141). If a GRS is positive, the subject carries more at-risk alleles. If a GRS is negative, the subject carries more beneficial alleles.

Figure 3. Receiver Operating Characteristic (ROC) Curve Analysis for the genetic risk score (GRS) in the *FAS* study population. Sensitivity and specificity were assessed using the logistic procedure in SAS v9.4.

Figure 4. Receiver Operating Characteristic (ROC) Curve Analysis for the genetic risk score (GRS) in the *FINGEN* study population. Sensitivity and specificity were assessed using the logistic procedure in SAS v9.4.