# Higher chylomicron remnants and LDL particle numbers associate with common *CD36* SNPs and DNA methylation sites that reduce *CD36* expression

Short title: CD36 SNPs influence postprandial lipids and LDL

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**Abbreviations:** CD36- cluster of differentiation 36; GOLDN- Genetics of Lipid Lowering Drugs and Diet Network; CpG-cytosine phosphate guanine; NMR- Nuclear magnetic resonance; MAF- minor allele frequency

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### Abstract

CD36 variants influence fasting lipids and risk of metabolic syndrome, but their impact on postprandial lipids, an independent risk factor for cardiovascular disease, is unclear. We determined effects of SNPs within a ~410-kb region encompassing CD36 and its proximal and distal promoters on chylomicron remnants (CM) and LDL particles at fasting, 3.5 and 6 hours following a high-fat meal (GOLDN study, n=1117). Five promoter variants associated with CM, four with delayed triglyceride clearance and five with LDL particle number. To assess mechanisms underlying the associations, we queried eQTL, DNA methylation and ChIP-seq datasets for adipose and heart, tissues that function in postprandial lipid clearance. Several SNPs that associated with higher serum lipids correlated with lower adipose and heart CD36 mRNA and aligned to active motifs for PPAR $\gamma$ , a major CD36 regulator. The SNPs also associated with DNA methylation sites that related to reduced CD36 mRNA and higher serum lipids but mixedmodel analyses indicated that the SNPs and methylation independently influence CD36 mRNA. The findings support contribution of CD36 SNPs that reduce adipose and heart CD36 RNA expression to inter-individual variability of postprandial lipid metabolism and document changes in CD36 DNA methylation that influence both CD36 expression and lipids.

### Introduction

Elevated fasting lipid levels have traditionally been associated with increased risk of heart disease, diabetes and stroke. More recently, similar and independent associations were found with levels of non-fasting or postprandial lipids (1, 2). Non-fasting measurements incorporate those of chylomicrons (CM) produced in response to dietary fat ingestion. The triglycerides (TG) in CM particles are hydrolyzed by lipoprotein lipase (LPL) in adipose, skeletal muscle and heart vascular beds to deliver free fatty acids (FFA) for cellular uptake facilitated by membrane CD36 (3). These tissues quantitatively contribute to chylomicron clearance in humans (4-6) and express high levels of CD36 at the level of both endothelial and parenchymal cells (7, 8). Elevated concentrations of postprandial CM remnants after an oral fat load were reported in four subjects with complete CD36 deficiency (9) suggesting that low CD36 might impair CM clearance and result in sustained high levels of circulating CM remnants (10, 11). Studies support a heritable component in determining postprandial lipid metabolism (12, 13). A number of candidate genes were implicated including apolipoproteins and LPL, but together they account for a modest fraction of postprandial lipid variability (14).

CD36 is a highly conserved membrane protein expressed on metabolic and immune cells that functions in cellular recognition and uptake of many lipid-related molecules (15, 16) including dietary long chain fatty acids and native or oxidized lipoproteins (17). In addition, CD36 functions in transducing cell type specific intracellular signaling that serves to regulate lipid utilization (17-19) or the immune response (20). Variants in the *CD36* gene have been shown to influence fasting lipids (21, 22) and risk of metabolic syndrome (MetS) (22). Evidence from CD36 null mouse models and CD36 deficient humans suggests that the protein regulates postprandial triglycerides and cholesterol metabolism. In mice, CD36 deficiency increases

cholesterol levels, impairs chylomicron secretion into the lymph and delays postprandial TG clearance (9, 23-25). A similar phenotype is found in CD36 deficient subjects who were reported to have increased chylomicron remnants and cholesterol levels (9, 26, 27). Despite the established role of CD36 in lipid absorption and CM metabolism, there is no information on whether polymorphisms in the *CD36* gene influence variability in postprandial lipid metabolism in humans. Such information is important considering the documented link between CM remnants and cardiovascular disease.

In this study, we determined the effects of common *CD36* variants (minor allele frequency 5-45%) on lipoprotein levels before and after ingestion of a high fat shake in the family based study of 1117 Caucasian participants, the Genetics of Lipid Lowering Drugs and Diet Network (GOLDN) (28). We identified several SNPs that influence CM remnants, TG response and LDL. The most significant SNPs overlap those previously associated with the MetS. Examination of the potential mechanisms underlying the observed effects highlighted transcriptional and epigenetic regulatory regions of the gene.

### **Materials and Methods**

**Study Population:** GOLDN participants of European ancestry were recruited from the NHLBI Family Heart Study. The data shown are from 1117 individuals (52% females); age:  $48.4 \pm 16.3$ , BMI:  $28.3 \pm 5.6$  and blood pressure: systolic  $115.3 \pm 16.6$ , diastolic  $68.1 \pm 9.3$  mmHg (Supplementary Table S1). Individuals  $\geq 19$  years old with fasting TG  $\leq 1500$  mg/dL, normal liver and kidney function were included while individuals with a history of pancreas, gallbladder or malabsorption diseases, use of lipid lowering drugs, insulin and warfarin were excluded, as

previously reported (29). Fasted (8 hours) participants ingested (within 15 minutes) a high fat milkshake (700 calories/m<sup>2</sup> body surface area, with 3% from protein, 14% from carbohydrate and 83% from fat). Blood was collected at fasting and after the meal at 3.5 and 6 hours for measuring lipid levels and lipoprotein profiling using nuclear magnetic resonance (NMR) (28). Blood collection at 3.5 and 6 hours is based on TG kinetics and designed to assess differences in lipid absorption and clearance, respectively (30-33).

**SNP selection and analysis:** Genotype data for 81 SNPs within a ~410-kb (chr7: 79,900,000-80,310,331) encompassing the *CD36* gene region met QC procedures and were used to impute 337 SNPs based on phased genotype data for HapMap CEU (release 22, Human Genome build 36) with minor allele frequencies (MAF)  $\geq$  5% (28). SNP genotyping was performed using the Affymetrix Human 6.0 platform (34). Linkage disequilibrium (LD) block structure was determined using the default method of Haploview for the Gabriel protocol (D' confidence interval 0.70-0.98) (35). Conventional and NMR derived lipid variables were transformed to render a normal distribution and adjusted for age, age<sup>2</sup> and sex within a stepwise regression. Lead SNPs were tested against lipid traits using the same linear mixed-effects model, accounting for familial relationship (36). TG uptake and clearance post meal were calculated from normalized growth curve models where uptake was based on time points 1 (fasting) and 2 (3.5 h post meal) and clearance on time point 3 (6 h post-meal) (28). Bonferroni corrections for multiple tests (p ≤ 0.006) were based on the major variables LDL, TG and 7 LD blocks across the *CD36* region (HapMap CEU, Version 4.2).

Metabolic phenotype, Transcript level, DNA methylation and genotypes in the MuTHER cohort: The MuTHER (Multiple Tissue Human Expression Resource) cohort includes subcutaneous adipose tissue collected from 856 females of European descent recruited from the

TwinsUK adult registry (37, 38). Adipose RNA was used for expression profiling (Illumina Human HT-12 V3 BeadChips) (38) while DNA was bisulfite converted (EZ-96 DNA Methylation Kit, Zymo Research) and profiled on the Illumina HumanMethylation450 array (Illumina Inc.) (37). Genotypes were generated using a combination of Illumina arrays (HumanHap300, HumanHap610Q, 1M Duo and 1.2MDuo 1M) and imputed with IMPUTE (v2) using two reference panels, P0 (HapMap2, rel 22, combined CEU+YRI+ASN panels) and P1 (610k+, including the combined HumanHap610k and 1M array). The SNPs were filtered at MAF >5% and IMPUTE info value of >0.8.

The MuTHER study analyzed DNA methylation and transcript levels for 210,984 methylation and 18,818 expression probes situated on or 1500bp upstream of 13,532 genes. For this study, only sites localized to *CD36* promoter region (chr7: 80,060,000-80,095,000) were examined. Association of DNA methylation and transcript levels with probabilities of imputed genotypes (MAF >5%, info >0.8) were tested in samples of related individuals using a two-step statistical approach implemented in the GenABEL/ProbABEL packages. Age, beadchip, BS conversion efficiency and BS-treated DNA input were included as cofactors in the DNA methylation analysis (metQTL, N<sub>total</sub>=603) (37) and age and experimental batch in the expression analysis (eQTL, N<sub>total</sub>=776) (38), respectively. The *cis*-metQTL analysis was limited to SNPs within 100-kb either side of probe location whereas the *cis*-eQTL analysis considered SNPs located within 1MB either side of transcription start or end site or within the gene body. False discovery rate (FDR) were calculated using the qvalue package 30 implemented in R2.11 26 (37, 38) where 1% FDR in the *cis*-metQTL, *cis*-eQTL and methylation vs. expression analysis corresponded to p<8.6E-4, p<5.0E-5 and p<2.6E-4, respectively.

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Associations between phenotypes or CD36 transcript levels with DNA methylation were tested with a linear mixed-effects model in R (39) using the lmer() function in the lme4 package (40) fitted by maximum-likelihood. The linear mixed-effects model was adjusted for both fixed (age, beadchip, BS conversion efficiency and BS-treated DNA input) and random effects (family relationship and zygosity). A likelihood ratio test was used to assess significance of the phenotype effect. The p-value of the phenotype effect in each model was calculated from the Chi-square distribution with 1 degree of freedom using -2log (likelihood ratio) as the test statistic.

### Results

**Common** *CD36* **SNPs influence chylomicron remnants and TG levels:** The GOLDN cohort includes relatively healthy, overweight to obese individuals (as indicated in Methods). The plasma lipid profile for this cohort before and after the meal is in Supplementary Table S1. To comprehensively examine effects of common *CD36* variants on postprandial lipids, we tested SNPs across the *CD36* locus encompassing all six alternative first exons (1A-1F) (41, 42) and the two 3'UTRs (short and long) (Fig. 1). SNP associations with CM levels were identified (Table 1 and Fig. 1) at 3.5h and 6h post meal. At 3.5 h, independent common SNPs near the distal promoter (1D in Fig. 1A) tagged by rs10081383, rs2048474 and rs13230620 associated with lower CM levels while CM levels were higher for the less frequent rs37920 (7%) (Table 1). At 6 h post meal, the same relationships were maintained for SNPs rs10081383 and rs37920 (Fig. 1B, Table 1). The effect of rs37920 (C allele) was much more significant at 6h suggesting that the major impact of the allele is to reduce clearance, a more substantial contributor to CM

levels at 6h post the meal. The rs10081383 (A-allele) and rs12535593 (intronic ~60kb downstream of 1D) SNPs associated with lower CM at 6h implying either reduced CM input or enhanced CM clearance. In addition, there was an effect in the 3'UTR at rs7755 (G, 46%) on increasing CM remnants (Fig. 1B) but this association was not significant after Bonferroni correction.

We then tested SNP associations with changes in TG levels post meal, using slope of growth curve modeling (43), which distinguishes between TG appearance (0-3.5h) and clearance (3.5 - 6.0 h) phases. As shown in Table 2, three common SNPs near the distal first exon 1D (rs304763, rs304798, rs304802) associated with slower TG appearance (or negative  $\beta$  estimate) and three other SNPs with delayed TG removal (rs37920, rs13228738, rs3211842). Of interest, rs37920 associated with higher CM remnants (Table 1), which is consistent with slow TG removal (Table 2). These findings support a role for the distal 1D region of *CD36* in the regulation of postprandial lipid metabolism.

*LDL*: Associations with higher LDL particle numbers at fasting and at the post meal times (Fig. 2, Table 3, Supplemental Table 2) were identified for SNPs in the central (near first exon 1C) and proximal (near 1B) promoters and in the longer 3'UTR. The most significant observations were with tag rs1761665 (C, 44%) and rs6970109 (A, 10%). Two additional SNPs sharing modest LD with rs1761665 showed similar effects. In the 3'UTR, rs7755 (A, 46%) (Table 3) previously linked to risk of MetS and stroke (21, 44) associated with higher LDL. As noted earlier it also associated with higher CM remnants at 6h post-meal but this did not survive the Bonferroni correction. Of note the associations shown for fasting LDL remained significant at 3.5 and 6h post meal (Supplementary Table S2) as would be expected.

Lipid-associated SNPs influence CD36 mRNA levels in adipose tissue and heart: To determine the potential functional relevance of the lipid-associated SNPs, we examined their influence on *CD36* level in adipose and heart, tissues that are highly active in lipid metabolism and where CD36 is abundant. Expression quantitative trait loci (eQTL) were available for adipose tissue from 776 healthy individuals of the MuTHER project (Multiple Tissue Human Expression Resource) (38). As shown in Table 4, analysis of the adipose tissue dataset documented strong influence of the SNPs on the most abundantly expressed CD36 transcript (1B long 3'UTR, NM 001001548) (41, 42). Alleles that related to increased CM remnants or higher LDL associated with lower adipose tissue CD36 mRNA (Table 4). Another independent genome-wide eQTL analysis performed with non-diseased human heart samples from 129 Western Europeans (45) showed similar negative relationships between the lipid-SNPs and CD36 mRNA (Table 4). These findings document strong effects of the SNPs on CD36 transcripts in heart and adipose tissue and suggest that expression in these tissues inversely relates to higher lipids consistent with the role of CD36 in enhancing lipid clearance and reducing circulating lipid levels.

SNPs align near sites for PPAR $\gamma$  and DNA methylation in adipose tissue *CD36*: Human adipocyte ChiP-seq data (46) showed that many of the associated promoter SNPs lie within close proximity of previously validated binding sites for PPAR $\gamma$ , a major *CD36* transcriptional regulator (Fig. 3). In addition, this region is characterized by active histone H3 lysine methylation and acetylation marks (46).

Previously, an adipose tissue epigenome study in MuTHER showed the most significant association with fasting TG levels at a CpG site (cg05917188) in the *CD36* promoter (47). In the

same dataset (37), we tested whether a significant relationship exists between the lipid-associated SNPs we identified and methylation using a multivariate linear regression model with age and BMI as covariates. Interestingly, the lipid-SNPs that negatively related to adipose CD36 level associated with DNA methylation at several CpG sites (cg25783969, cg21055948, cg19096849 and cg18508525, cg26138637, cg06601993) (Table 5) within the central promoter region. Hypomethylation of four of these sites strongly related to CD36 mRNA levels (Table 6). To gain a better perspective of the contribution of altered methylation to the SNP effects on CD36 mRNA levels, a mixed-linear model for effects of the two most significant eQTLs, rs1761665 (promoter) or rs7755 (3'UTR) was adjusted for methylation (Supplementary Table S3). Comparison of the adjusted beta estimates shown in Supplementary Table S3 to those shown in Table 4, where associations are unadjusted demonstrates that adjustment for methylation does not significantly attenuate the associations between the SNPs and the most abundant CD36 transcript (1B with long 3'UTR). This indicated that the effect of the SNPs on CD36 is independent of methylation at the above sites and is more likely to involve transcriptional regulation of the gene by PPAR $\gamma$  or posttranslational regulatory sites in the 3'UTR (rs7755). We then examined whether methylation also relates to higher lipid levels. As shown in Table 7, DNA methylation correlated with higher TG and LDL, suggesting that it independently impacts lipid metabolism through its influence on CD36 level (Table 6).

### Discussion

Complex interactions between diet, lifestyle, and genetic factors can lead to sustained elevation of postprandial lipids, which independently increases the risk of cardiovascular disease

(1, 2). Earlier genetic studies linked *CD36* to fasting lipids (21, 48), MetS (22, 49-51) and stroke risk (44). However, information related to the role of the gene in postprandial lipid metabolism, which plays an important role in etiology of the MetS and cardiovascular disease remains limited. This study is the first comprehensive examination of the impact of CD36 variants on postprandial lipids. We present findings implicating common promoter SNPs in inter-individual variability of the CM remnant response after a fatty meal. We also identify SNPs that associate with LDL particle number at both fasting and two postprandial time points. We document that the identified lipid SNPs, which align close to validated PPAR $\gamma$  sites, associate with reduced adipose and heart CD36 and higher lipid traits supporting the role of CD36 in enhancing lipid clearance. In addition, we provide evidence that higher serum lipids associate with altered DNA methylation and reduced *CD36* mRNA.

The most significant promoter SNPs identified to influence CM levels and TG clearance (rs1931694, rs13228738, rs3211842) are in strong LD with variants previously associated with MetS or coronary artery disease, CAD (22, 48, 49). Similarly, the 3'UTR variant rs7755 identified to associate with LDL and to trend with higher CM remnants has been linked to diabetes-associated CAD (48) stroke (44) as well as MetS (22). Overall our findings support an important role of CD36 in the handling of dietary lipids in humans and are consistent with data in CD36 null mice showing defective CM metabolism and persistently high levels of CM remnants (9, 23-25). Elevated CM remnant levels were reported in small cohorts of CD36 deficient subjects (9, 26, 27). Our study indicates that common *CD36* alleles (5-45% frequency) contribute to variability of the postprandial response and possibly to risk of diet-induced metabolic abnormalities. GOLDN is the largest cohort to-date in which postprandial lipid measures are available and ideally our findings will need replication in other large studies as they

become available.

We identified five CD36 SNPs to associate with higher LDL while previous GWAS did not link LDL to the CD36 locus. Several considerations might explain the discrepancy. The GOLDN family-based study consists of relatively healthy overweight to obese Caucasians who underwent a washout period to neutralize effects of lipid-lowering medications. Earlier GWAS included cohorts with a wide range of ages, metabolic disease traits, medications and BMI (52). We queried available Global Lipid Genetics (GLG) Consortium datasets using LocusZoom and found that most cohorts did not include sufficient SNP coverage of the CD36 promoter region comparable to this study. In one of the GLG studies (53) with similar coverage and that included SNPs in LD with those we genotyped, we found modest raw associations (p<0.05) with LDL cholesterol. Several of the identified variants, rs799979, rs646722, rs6949840 rs2030711, and rs17154155, are in strong LD with rs1722507 and rs1761665 in this study. In addition, our study measured LDL particle number as opposed to LDL cholesterol obtained by the Friedwald calculation. While both measures predict cardiovascular risk, they do not always correlate (1, 54). In support of our findings, are the reports of higher cholesterol levels in partially (55) or completely CD36 deficient individuals (9) and the higher LDL levels and abnormal cholesterol absorption of CD36 null mice (24, 56). In addition, a recent linkage analysis of LDL particle response to fenofibrate treatment in GOLDN reported a peak encompassing the CD36 locus (57). Thus, the LDL associations we identified and their potential relationship to disease risk warrant further investigation.

Our findings highlight regulatory regions in the *CD36* gene: SNPs near the distal promoter (designated 1D), which is almost exclusive to endothelial cells (42), associated with TG uptake/clearance and CM remnants. The central promoter region had signals for LDL and

TG clearance, and contained the validated PPAR $\gamma$  sites and the lipid-associated DNA methylation sites. Finally, the long 3'UTR had signals for LDL and contained predicted miRNA and RNA editing sites.

We showed that SNPs that associated with higher CM remnants and LDL reduced adipose and heart *CD36*. The rs1761665 SNP in particular is in strong LD with promoter SNPs previously shown to influence monocyte and platelet CD36 (58, 59). Several of the promoter SNPs we identified here, lie in close proximity to validated binding sites for PPAR $\gamma$  and to methylation sites that associate with *CD36* mRNA levels (Figure 3). Further analyses involving the two most significant eQTLs (rs1761665 and rs7755) for CD36 indicated that the effect of these SNPs is independent of methylation and is more likely to involve altered transcriptional regulation by PPAR $\gamma$  (rs1761665) or posttranscriptional effects involving the 3'UTR (rs7755). Although changes in methylation do not account for the SNP effect on CD36 level, our data suggest that they influence lipid levels. Methylation might contribute to the metabolic and dietary regulation of the *CD36* gene. For example, fatty acids were reported to alter global DNA methylation and to participate in shaping metabolic disease related methylomes (60).

In summary, our findings link common *CD36* SNPs that reduce adipose and heart *CD36* levels to higher CM remnants and LDL in humans. These SNPs would contribute to individual differences in the handling of dietary lipid and in susceptibility to diet induced metabolic abnormalities. Finally, factors that alter methylation of the *CD36* gene impact lipid levels and potentially disease risk.

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# Disclosures

None declared.

### References

1. Mora, S., N. Rifai, J. E. Buring, and P. M. Ridker. 2008. Fasting compared with nonfasting lipids and apolipoproteins for predicting incident cardiovascular events. *Circulation* **118**: 993-1001.

2. Nordestgaard, B. G., and A. Varbo. 2014. Triglycerides and cardiovascular disease. *Lancet* **384**: 626-635.

3. Goldberg, I. J., R. H. Eckel, and N. A. Abumrad. 2009. Regulation of fatty acid uptake into tissues: lipoprotein lipase- and CD36-mediated pathways. *Journal of lipid research* **50 Suppl**: S86-90.

Bickerton, A. S., R. Roberts, B. A. Fielding, L. Hodson, E. E. Blaak, A. J. Wagenmakers,
M. Gilbert, F. Karpe, and K. N. Frayn. 2007. Preferential uptake of dietary Fatty acids in adipose
tissue and muscle in the postprandial period. *Diabetes* 56: 168-176.

5. Karpe, F., A. S. Bickerton, L. Hodson, B. A. Fielding, G. D. Tan, and K. N. Frayn. 2007. Removal of triacylglycerols from chylomicrons and VLDL by capillary beds: the basis of lipoprotein remnant formation. *Biochemical Society transactions* **35**: 472-476.

6. Potts, J. L., S. W. Coppack, R. M. Fisher, S. M. Humphreys, G. F. Gibbons, and K. N. Frayn. 1995. Impaired postprandial clearance of triacylglycerol-rich lipoproteins in adipose tissue in obese subjects. *Am J Physiol* **268**: E588-594.

7. Greenwalt, D. E., S. H. Scheck, and T. Rhinehart-Jones. 1995. Heart CD36 expression is increased in murine models of diabetes and in mice fed a high fat diet. *The Journal of clinical investigation* **96**: 1382-1388.

8. Abumrad, N. A., and I. J. Goldberg. 2016. CD36 actions in the heart: Lipids, calcium, inflammation, repair and more? *Biochimica et biophysica acta*.

JOURNAL OF LIPID RESEARCH

9. Masuda, D., K. Hirano, H. Oku, J. C. Sandoval, R. Kawase, M. Yuasa-Kawase, Y. Yamashita, M. Takada, K. Tsubakio-Yamamoto, Y. Tochino, M. Koseki, F. Matsuura, M. Nishida, T. Kawamoto, M. Ishigami, M. Hori, I. Shimomura, and S. Yamashita. 2009. Chylomicron remnants are increased in the postprandial state in CD36 deficiency. *Journal of lipid research* **50**: 999-1011.

10. Adiels, M., N. Matikainen, J. Westerbacka, S. Soderlund, T. Larsson, S. O. Olofsson, J. Boren, and M. R. Taskinen. 2012. Postprandial accumulation of chylomicrons and chylomicron remnants is determined by the clearance capacity. *Atherosclerosis* **222**: 222-228.

11. Frayn, K., S. Bernard, K. Spalding, and P. Arner. 2012. Adipocyte triglyceride turnover is independently associated with atherogenic dyslipidemia. *Journal of the American Heart Association* **1**: e003467.

12. Irvin, M. R., D. Zhi, S. Aslibekyan, S. A. Claas, D. M. Absher, J. M. Ordovas, H. K. Tiwari, S. Watkins, and D. K. Arnett. 2014. Genomics of post-prandial lipidomic phenotypes in the Genetics of Lipid lowering Drugs and Diet Network (GOLDN) study. *PloS one* **9**: e99509.

13. Uiterwaal, C. S., D. E. Grobbee, J. C. Witteman, W. A. van Stiphout, X. H. Krauss, L. M. Havekes, A. M. de Bruijn, A. van Tol, and A. Hofman. 1994. Postprandial triglyceride response in young adult men and familial risk for coronary atherosclerosis. *Annals of internal medicine* **121**: 576-583.

14. Perez-Martinez, P., J. Lopez-Miranda, F. Perez-Jimenez, and J. M. Ordovas. 2008. Influence of genetic factors in the modulation of postprandial lipemia. *Atherosclerosis. Supplements* **9**: 49-55. Hoebe, K., P. Georgel, S. Rutschmann, X. Du, S. Mudd, K. Crozat, S. Sovath, L. Shamel, T. Hartung, U. Zahringer, and B. Beutler. 2005. CD36 is a sensor of diacylglycerides. *Nature* 433: 523-527.

Abumrad, N. A., and N. O. Davidson. 2012. Role of the gut in lipid homeostasis.
 *Physiological reviews* 92: 1061-1085.

17. Endemann, G., L. W. Stanton, K. S. Madden, C. M. Bryant, R. T. White, and A. A. Protter. 1993. CD36 is a receptor for oxidized low density lipoprotein. *The Journal of biological chemistry* **268**: 11811-11816.

18. Hames, K. C., A. Vella, B. J. Kemp, and M. D. Jensen. 2014. Free fatty acid uptake in humans with CD36 deficiency. *Diabetes* **63**: 3606-3614.

19. Pepino, M. Y., O. Kuda, D. Samovski, and N. A. Abumrad. 2014. Structure-function of CD36 and importance of fatty acid signal transduction in fat metabolism. *Annual review of nutrition* **34**: 281-303.

20. Stewart, C. R., L. M. Stuart, K. Wilkinson, J. M. van Gils, J. Deng, A. Halle, K. J. Rayner, L. Boyer, R. Zhong, W. A. Frazier, A. Lacy-Hulbert, J. El Khoury, D. T. Golenbock, and K. J. Moore. 2010. CD36 ligands promote sterile inflammation through assembly of a Toll-like receptor 4 and 6 heterodimer. *Nature immunology* **11**: 155-161.

21. Love-Gregory, L., and N. A. Abumrad. 2011. CD36 genetics and the metabolic complications of obesity. *Current opinion in clinical nutrition and metabolic care* **14**: 527-534.

22. Love-Gregory, L., R. Sherva, L. Sun, J. Wasson, T. Schappe, A. Doria, D. C. Rao, S. C. Hunt, S. Klein, R. J. Neuman, M. A. Permutt, and N. A. Abumrad. 2008. Variants in the CD36 gene associate with the metabolic syndrome and high-density lipoprotein cholesterol. *Human molecular genetics* **17**: 1695-1704.

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Downloaded from www.jlr.org at Queen Mary, University of London, on June 2, 2017

23. Tran, T. T., H. Poirier, L. Clement, F. Nassir, M. M. Pelsers, V. Petit, P. Degrace, M. C. Monnot, J. F. Glatz, N. A. Abumrad, P. Besnard, and I. Niot. 2011. Luminal lipid regulates CD36 levels and downstream signaling to stimulate chylomicron synthesis. *The Journal of biological chemistry* **286**: 25201-25210.

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24. Nauli, A. M., F. Nassir, S. Zheng, Q. Yang, C. M. Lo, S. B. Vonlehmden, D. Lee, R. J. Jandacek, N. A. Abumrad, and P. Tso. 2006. CD36 is important for chylomicron formation and secretion and may mediate cholesterol uptake in the proximal intestine. *Gastroenterology* **131**: 1197-1207.

25. Drover, V. A., M. Ajmal, F. Nassir, N. O. Davidson, A. M. Nauli, D. Sahoo, P. Tso, and N. A. Abumrad. 2005. CD36 deficiency impairs intestinal lipid secretion and clearance of chylomicrons from the blood. *The Journal of clinical investigation* **115**: 1290-1297.

26. Yamashita, S., K. Hirano, T. Kuwasako, M. Janabi, Y. Toyama, M. Ishigami, and N. Sakai. 2007. Physiological and pathological roles of a multi-ligand receptor CD36 in atherogenesis; insights from CD36-deficient patients. *Molecular and cellular biochemistry* **299**: 19-22.

27. Kuwasako, T., K. Hirano, N. Sakai, M. Ishigami, H. Hiraoka, M. J. Yakub, K. Yamauchi-Takihara, S. Yamashita, and Y. Matsuzawa. 2003. Lipoprotein abnormalities in human genetic CD36 deficiency associated with insulin resistance and abnormal fatty acid metabolism. *Diabetes care* **26**: 1647-1648.

28. Wojczynski, M. K., S. P. Glasser, A. Oberman, E. K. Kabagambe, P. N. Hopkins, M. Y. Tsai, R. J. Straka, J. M. Ordovas, and D. K. Arnett. 2011. High-fat meal effect on LDL, HDL, and VLDL particle size and number in the Genetics of Lipid-Lowering Drugs and Diet Network (GOLDN): an interventional study. *Lipids in health and disease* **10**: 181.

29. Corella, D., D. K. Arnett, M. Y. Tsai, E. K. Kabagambe, J. M. Peacock, J. E. Hixson, R. J. Straka, M. Province, C. Q. Lai, L. D. Parnell, I. Borecki, and J. M. Ordovas. 2007. The - 256T>C polymorphism in the apolipoprotein A-II gene promoter is associated with body mass index and food intake in the genetics of lipid lowering drugs and diet network study. *Clinical chemistry* **53**: 1144-1152.

ASBMB

JOURNAL OF LIPID RESEARCH

30. Patsch, J. R., G. Miesenbock, T. Hopferwieser, V. Muhlberger, E. Knapp, J. K. Dunn, A.
M. Gotto, Jr., and W. Patsch. 1992. Relation of triglyceride metabolism and coronary artery disease. Studies in the postprandial state. *Arteriosclerosis and thrombosis : a journal of vascular biology / American Heart Association* 12: 1336-1345.

31. Dubois, C., M. Armand, V. Azais-Braesco, H. Portugal, A. M. Pauli, P. M. Bernard, C. Latge, H. Lafont, P. Borel, and D. Lairon. 1994. Effects of moderate amounts of emulsified dietary fat on postprandial lipemia and lipoproteins in normolipidemic adults. *The American journal of clinical nutrition* **60**: 374-382.

32. Lopez-Miranda, J., G. Cruz, P. Gomez, C. Marin, E. Paz, P. Perez-Martinez, F. J. Fuentes, J. M. Ordovas, and F. Perez-Jimenez. 2004. The influence of lipoprotein lipase gene variation on postprandial lipoprotein metabolism. *The Journal of clinical endocrinology and metabolism* **89**: 4721-4728.

33. Williams, C. M. 1997. Postprandial lipid metabolism: effects of dietary fatty acids. *The Proceedings of the Nutrition Society* **56**: 679-692.

34. Irvin, M. R., E. K. Kabagambe, H. K. Tiwari, L. D. Parnell, R. J. Straka, M. Tsai, J. M. Ordovas, and D. K. Arnett. 2010. Apolipoprotein E polymorphisms and postprandial triglyceridemia before and after fenofibrate treatment in the Genetics of Lipid Lowering and Diet Network (GOLDN) Study. *Circulation. Cardiovascular genetics* **3**: 462-467.

35. Gabriel, S. B., S. F. Schaffner, H. Nguyen, J. M. Moore, J. Roy, B. Blumenstiel, J. Higgins, M. DeFelice, A. Lochner, M. Faggart, S. N. Liu-Cordero, C. Rotimi, A. Adeyemo, R. Cooper, R. Ward, E. S. Lander, M. J. Daly, and D. Altshuler. 2002. The structure of haplotype blocks in the human genome. *Science* **296**: 2225-2229.

36. Kenward, M. G., and J. H. Roger. 1997. Small sample inference for fixed effects from restricted maximum likelihood. *Biometrics* **53**: 983-997.

**BASBMB** 

JOURNAL OF LIPID RESEARCH

37. Grundberg, E., E. Meduri, J. K. Sandling, A. K. Hedman, S. Keildson, A. Buil, S. Busche, W. Yuan, J. Nisbet, M. Sekowska, A. Wilk, A. Barrett, K. S. Small, B. Ge, M. Caron, S. Y. Shin, C. Multiple Tissue Human Expression Resource, M. Lathrop, E. T. Dermitzakis, M. I. McCarthy, T. D. Spector, J. T. Bell, and P. Deloukas. 2013. Global analysis of DNA methylation variation in adipose tissue from twins reveals links to disease-associated variants in distal regulatory elements. *American journal of human genetics* **93**: 876-890.

38. Grundberg, E., K. S. Small, A. K. Hedman, A. C. Nica, A. Buil, S. Keildson, J. T. Bell, T. P. Yang, E. Meduri, A. Barrett, J. Nisbett, M. Sekowska, A. Wilk, S. Y. Shin, D. Glass, M. Travers, J. L. Min, S. Ring, K. Ho, G. Thorleifsson, A. Kong, U. Thorsteindottir, C. Ainali, A. S. Dimas, N. Hassanali, C. Ingle, D. Knowles, M. Krestyaninova, C. E. Lowe, P. Di Meglio, S. B. Montgomery, L. Parts, S. Potter, G. Surdulescu, L. Tsaprouni, S. Tsoka, V. Bataille, R. Durbin, F. O. Nestle, S. O'Rahilly, N. Soranzo, C. M. Lindgren, K. T. Zondervan, K. R. Ahmadi, E. E. Schadt, K. Stefansson, G. D. Smith, M. I. McCarthy, P. Deloukas, E. T. Dermitzakis, T. D. Spector, and C. Multiple Tissue Human Expression Resource. 2012. Mapping cis- and transregulatory effects across multiple tissues in twins. *Nature genetics* 44: 1084-1089.

39. Team, R. D. C. 2011. R: A Language and Environment for Statistical Computing. *In* R Foundation for Statistical Computing, Vienna, Austria.

resj 44. Del Lau Kel Yan M

JOURNAL OF LIPID RESEARCH

40. Bates, D., Maechler, M., Bolker, B. 2011. lme4: linear mixed-effects models using S4 classes.

41. Andersen, M., B. Lenhard, C. Whatling, P. Eriksson, and J. Odeberg. 2006. Alternative promoter usage of the membrane glycoprotein CD36. *BMC Mol Biol* **7**: 8.

42. Pietka, T. A., T. Schappe, C. Conte, E. Fabbrini, B. W. Patterson, S. Klein, N. A. Abumrad, and L. Love-Gregory. 2014. Adipose and muscle tissue profile of CD36 transcripts in obese subjects highlights the role of CD36 in fatty acid homeostasis and insulin resistance. *Diabetes care* **37**: 1990-1997.

43. Wojczynski, M. K., G. Gao, I. Borecki, P. N. Hopkins, L. Parnell, C. Q. Lai, J. M. Ordovas, B. H. Chung, and D. K. Arnett. 2010. Apolipoprotein B genetic variants modify the response to fenofibrate: a GOLDN study. *Journal of lipid research* **51**: 3316-3323.

Ikram, M. A., S. Seshadri, J. C. Bis, M. Fornage, A. L. DeStefano, Y. S. Aulchenko, S. Debette, T. Lumley, A. R. Folsom, E. G. van den Herik, M. J. Bos, A. Beiser, M. Cushman, L. J. Launer, E. Shahar, M. Struchalin, Y. Du, N. L. Glazer, W. D. Rosamond, F. Rivadeneira, M. Kelly-Hayes, O. L. Lopez, J. Coresh, A. Hofman, C. DeCarli, S. R. Heckbert, P. J. Koudstaal, Q. Yang, N. L. Smith, C. S. Kase, K. Rice, T. Haritunians, G. Roks, P. L. de Kort, K. D. Taylor, L. M. de Lau, B. A. Oostra, A. G. Uitterlinden, J. I. Rotter, E. Boerwinkle, B. M. Psaty, T. H. Mosley, C. M. van Duijn, M. M. Breteler, W. T. Longstreth, Jr., and P. A. Wolf. 2009. Genomewide association studies of stroke. *The New England journal of medicine* 360: 1718-1728.

45. Koopmann, T. T., M. E. Adriaens, P. D. Moerland, R. F. Marsman, M. L. Westerveld, S. Lal, T. Zhang, C. Q. Simmons, I. Baczko, C. dos Remedios, N. H. Bishopric, A. Varro, A. L.

George, Jr., E. M. Lodder, and C. R. Bezzina. 2014. Genome-wide identification of expression quantitative trait loci (eQTLs) in human heart. *PloS one* **9**: e97380.

46. Mikkelsen, T. S., Z. Xu, X. Zhang, L. Wang, J. M. Gimble, E. S. Lander, and E. D. Rosen. 2010. Comparative epigenomic analysis of murine and human adipogenesis. *Cell* **143**: 156-169.

47. Allum, F., X. Shao, F. Guenard, M. M. Simon, S. Busche, M. Caron, J. Lambourne, J. Lessard, K. Tandre, A. K. Hedman, T. Kwan, B. Ge, C. Multiple Tissue Human Expression Resource, L. Ronnblom, M. I. McCarthy, P. Deloukas, T. Richmond, D. Burgess, T. D. Spector, A. Tchernof, S. Marceau, M. Lathrop, M. C. Vohl, T. Pastinen, and E. Grundberg. 2015. Characterization of functional methylomes by next-generation capture sequencing identifies novel disease-associated variants. *Nature communications* **6**: 7211.

48. Ma, X., S. Bacci, W. Mlynarski, L. Gottardo, T. Soccio, C. Menzaghi, E. Iori, R. A. Lager, A. R. Shroff, E. V. Gervino, R. W. Nesto, M. T. Johnstone, N. A. Abumrad, A. Avogaro, V. Trischitta, and A. Doria. 2004. A common haplotype at the CD36 locus is associated with high free fatty acid levels and increased cardiovascular risk in Caucasians. *Human molecular genetics* **13**: 2197-2205.

49. Farook, V. S., S. Puppala, J. Schneider, S. P. Fowler, G. Chittoor, T. D. Dyer, H. Allayee,
S. A. Cole, R. Arya, M. H. Black, J. E. Curran, L. Almasy, T. A. Buchanan, C. P. Jenkinson, D.
M. Lehman, R. M. Watanabe, J. Blangero, and R. Duggirala. 2012. Metabolic syndrome is
linked to chromosome 7q21 and associated with genetic variants in CD36 and GNAT3 in
Mexican Americans. *Obesity* 20: 2083-2092.

Coram, M. A., Q. Duan, T. J. Hoffmann, T. Thornton, J. W. Knowles, N. A. Johnson, H.
 M. Ochs-Balcom, T. A. Donlon, L. W. Martin, C. B. Eaton, J. G. Robinson, N. J. Risch, X. Zhu,

ASBMB

C. Kooperberg, Y. Li, A. P. Reiner, and H. Tang. 2013. Genome-wide characterization of shared and distinct genetic components that influence blood lipid levels in ethnically diverse human populations. *American journal of human genetics* **92**: 904-916.

51. Elbers, C. C., Y. Guo, V. Tragante, E. P. van Iperen, M. B. Lanktree, B. A. Castillo, F. Chen, L. R. Yanek, M. K. Wojczynski, Y. R. Li, B. Ferwerda, C. M. Ballantyne, S. G. Buxbaum, Y. D. Chen, W. M. Chen, L. A. Cupples, M. Cushman, Y. Duan, D. Duggan, M. K. Evans, J. K. Fernandes, M. Fornage, M. Garcia, W. T. Garvey, N. Glazer, F. Gomez, T. B. Harris, I. Halder, V. J. Howard, M. F. Keller, M. I. Kamboh, C. Kooperberg, S. B. Kritchevsky, A. LaCroix, K. Liu, Y. Liu, K. Musunuru, A. B. Newman, N. C. Onland-Moret, J. Ordovas, I. Peter, W. Post, S. Redline, S. E. Reis, R. Saxena, P. J. Schreiner, K. A. Volcik, X. Wang, S. Yusuf, A. B. Zonderland, S. S. Anand, D. M. Becker, B. Psaty, D. J. Rader, A. P. Reiner, S. S. Rich, J. I. Rotter, M. M. Sale, M. Y. Tsai, I. B. Borecki, R. A. Hegele, S. Kathiresan, M. A. Nalls, H. A. Taylor, Jr., H. Hakonarson, S. Sivapalaratnam, F. W. Asselbergs, F. Drenos, J. G. Wilson, and B. J. Keating. 2012. Gene-centric meta-analysis of lipid traits in African, East Asian and Hispanic populations. *PloS one* 7: e50198.

52. Global Lipids Genetics, C., C. J. Willer, E. M. Schmidt, S. Sengupta, G. M. Peloso, S. Gustafsson, S. Kanoni, A. Ganna, J. Chen, M. L. Buchkovich, S. Mora, J. S. Beckmann, J. L. Bragg-Gresham, H. Y. Chang, A. Demirkan, H. M. Den Hertog, R. Do, L. A. Donnelly, G. B. Ehret, T. Esko, M. F. Feitosa, T. Ferreira, K. Fischer, P. Fontanillas, R. M. Fraser, D. F. Freitag, D. Gurdasani, K. Heikkila, E. Hypponen, A. Isaacs, A. U. Jackson, A. Johansson, T. Johnson, M. Kaakinen, J. Kettunen, M. E. Kleber, X. Li, J. Luan, L. P. Lyytikainen, P. K. Magnusson, M. Mangino, E. Mihailov, M. E. Montasser, M. Muller-Nurasyid, I. M. Nolte, J. R. O'Connell, C. D. Palmer, M. Perola, A. K. Petersen, S. Sanna, R. Saxena, S. K. Service, S. Shah, D. Shungin, C.

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Sidore, C. Song, R. J. Strawbridge, I. Surakka, T. Tanaka, T. M. Teslovich, G. Thorleifsson, E. G. Van den Herik, B. F. Voight, K. A. Volcik, L. L. Waite, A. Wong, Y. Wu, W. Zhang, D. Absher, G. Asiki, I. Barroso, L. F. Been, J. L. Bolton, L. L. Bonnycastle, P. Brambilla, M. S. Burnett, G. Cesana, M. Dimitriou, A. S. Doney, A. Doring, P. Elliott, S. E. Epstein, G. I. Evjolfsson, B. Gigante, M. O. Goodarzi, H. Grallert, M. L. Gravito, C. J. Groves, G. Hallmans, A. L. Hartikainen, C. Hayward, D. Hernandez, A. A. Hicks, H. Holm, Y. J. Hung, T. Illig, M. R. Jones, P. Kaleebu, J. J. Kastelein, K. T. Khaw, E. Kim, N. Klopp, P. Komulainen, M. Kumari, C. Langenberg, T. Lehtimaki, S. Y. Lin, J. Lindstrom, R. J. Loos, F. Mach, W. L. McArdle, C. Meisinger, B. D. Mitchell, G. Muller, R. Nagaraja, N. Narisu, T. V. Nieminen, R. N. Nsubuga, I. Olafsson, K. K. Ong, A. Palotie, T. Papamarkou, C. Pomilla, A. Pouta, D. J. Rader, M. P. Reilly, P. M. Ridker, F. Rivadeneira, I. Rudan, A. Ruokonen, N. Samani, H. Scharnagl, J. Seeley, K. Silander, A. Stancakova, K. Stirrups, A. J. Swift, L. Tiret, A. G. Uitterlinden, L. J. van Pelt, S. Vedantam, N. Wainwright, C. Wijmenga, S. H. Wild, G. Willemsen, T. Wilsgaard, J. F. Wilson, E. H. Young, J. H. Zhao, L. S. Adair, D. Arveiler, T. L. Assimes, S. Bandinelli, F. Bennett, M. Bochud, B. O. Boehm, D. I. Boomsma, I. B. Borecki, S. R. Bornstein, P. Bovet, M. Burnier, H. Campbell, A. Chakravarti, J. C. Chambers, Y. D. Chen, F. S. Collins, R. S. Cooper, J. Danesh, G. Dedoussis, U. de Faire, A. B. Feranil, J. Ferrieres, L. Ferrucci, N. B. Freimer, C. Gieger, L. C. Groop, V. Gudnason, U. Gyllensten, A. Hamsten, T. B. Harris, A. Hingorani, J. N. Hirschhorn, A. Hofman, G. K. Hovingh, C. A. Hsiung, S. E. Humphries, S. C. Hunt, K. Hveem, C. Iribarren, M. R. Jarvelin, A. Jula, M. Kahonen, J. Kaprio, A. Kesaniemi, M. Kivimaki, J. S. Kooner, P. J. Koudstaal, R. M. Krauss, D. Kuh, J. Kuusisto, K. O. Kyvik, M. Laakso, T. A. Lakka, L. Lind, C. M. Lindgren, N. G. Martin, W. Marz, M. I. McCarthy, C. A. McKenzie, P. Meneton, A. Metspalu, L. Moilanen, A. D. Morris, P. B. Munroe, I. Njolstad, N. L. Pedersen, C. Power, P. P.

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Pramstaller, J. F. Price, B. M. Psaty, T. Quertermous, R. Rauramaa, D. Saleheen, V. Salomaa, D.
K. Sanghera, J. Saramies, P. E. Schwarz, W. H. Sheu, A. R. Shuldiner, A. Siegbahn, T. D.
Spector, K. Stefansson, D. P. Strachan, B. O. Tayo, E. Tremoli, J. Tuomilehto, M. Uusitupa, C.
M. van Duijn, P. Vollenweider, L. Wallentin, N. J. Wareham, J. B. Whitfield, B. H.
Wolffenbuttel, J. M. Ordovas, E. Boerwinkle, C. N. Palmer, U. Thorsteinsdottir, D. I. Chasman,
J. I. Rotter, P. W. Franks, S. Ripatti, L. A. Cupples, M. S. Sandhu, S. S. Rich, M. Boehnke, P.
Deloukas, S. Kathiresan, K. L. Mohlke, E. Ingelsson, and G. R. Abecasis. 2013. Discovery and

Kathiresan, S., C. J. Willer, G. M. Peloso, S. Demissie, K. Musunuru, E. E. Schadt, L. Kaplan, D. Bennett, Y. Li, T. Tanaka, B. F. Voight, L. L. Bonnycastle, A. U. Jackson, G. Crawford, A. Surti, C. Guiducci, N. P. Burtt, S. Parish, R. Clarke, D. Zelenika, K. A. Kubalanza, M. A. Morken, L. J. Scott, H. M. Stringham, P. Galan, A. J. Swift, J. Kuusisto, R. N. Bergman, J. Sundvall, M. Laakso, L. Ferrucci, P. Scheet, S. Sanna, M. Uda, Q. Yang, K. L. Lunetta, J. Dupuis, P. I. de Bakker, C. J. O'Donnell, J. C. Chambers, J. S. Kooner, S. Hercberg, P. Meneton, E. G. Lakatta, A. Scuteri, D. Schlessinger, J. Tuomilehto, F. S. Collins, L. Groop, D. Altshuler, R. Collins, G. M. Lathrop, O. Melander, V. Salomaa, L. Peltonen, M. Orho-Melander, J. M. Ordovas, M. Boehnke, G. R. Abecasis, K. L. Mohlke, and L. A. Cupples. 2009. Common variants at 30 loci contribute to polygenic dyslipidemia. *Nature genetics* 41: 56-65.

54. Otvos, J. D., S. Mora, I. Shalaurova, P. Greenland, R. H. Mackey, and D. C. Goff, Jr. 2011. Clinical implications of discordance between low-density lipoprotein cholesterol and particle number. *Journal of clinical lipidology* **5**: 105-113.

55. Yanai, H., H. Chiba, M. Morimoto, K. Abe, H. Fujiwara, H. Fuda, S. P. Hui, Y. Takahashi, H. Akita, G. A. Jamieson, K. Kobayashi, and K. Matsuno. 2000. Human CD36

ASBMB

deficiency is associated with elevation in low-density lipoprotein-cholesterol. *American journal of medical genetics* **93**: 299-304.

Febbraio, M., N. A. Abumrad, D. P. Hajjar, K. Sharma, W. Cheng, S. F. Pearce, and R.
 L. Silverstein. 1999. A null mutation in murine CD36 reveals an important role in fatty acid and lipoprotein metabolism. *The Journal of biological chemistry* 274: 19055-19062.

57. Hidalgo, B., S. Aslibekyan, H. W. Wiener, M. R. Irvin, R. J. Straka, I. B. Borecki, H. K. Tiwari, M. Y. Tsai, P. N. Hopkins, J. M. Ordovas, and D. K. Arnett. 2015. A family-specific linkage analysis of blood lipid response to fenofibrate in the Genetics of Lipid Lowering Drug and Diet Network. *Pharmacogenetics and genomics* **25**: 511-514.

58. Ghosh, A., G. Murugesan, K. Chen, L. Zhang, Q. Wang, M. Febbraio, R. M. Anselmo, K. Marchant, J. Barnard, and R. L. Silverstein. 2011. Platelet CD36 surface expression levels affect functional responses to oxidized LDL and are associated with inheritance of specific genetic polymorphisms. *Blood* **117**: 6355-6366.

59. Love-Gregory, L., R. Sherva, T. Schappe, J. S. Qi, J. McCrea, S. Klein, M. A. Connelly, and N. A. Abumrad. 2011. Common CD36 SNPs reduce protein expression and may contribute to a protective atherogenic profile. *Human molecular genetics* **20**: 193-201.

60. Silva-Martinez, G. A., D. Rodriguez-Rios, Y. Alvarado-Caudillo, A. Vaquero, M. Esteller, F. J. Carmona, S. Moran, F. C. Nielsen, M. Wickstrom-Lindholm, K. Wrobel, K. Wrobel, G. Barbosa-Sabanero, S. Zaina, and G. Lund. 2016. Arachidonic and oleic acid exert distinct effects on the DNA methylome. *Epigenetics* **11**: 321-334.

SNP	Effect Allele	MAF	β(se)	* <i>p-</i> value		
<u>3.5 h post-meal</u>						
rs10081383	$G/\underline{A}$	0.24	-0.18 (0.06)	0.0035		
rs37920	T/ <u>C</u>	0.07	0.40 (0.14)	0.0034		
rs2048474	A/ <u>G</u>	0.32	-0.19 (0.06)	0.0011		
rs13230620	G/ <u>C</u>	0.28	-0.18 (0.06)	0.0032		
6.0 h post-meal						
rs10081383	G/ <u>A</u>	0.24	-0.18 (0.06)	0.0005		
rs37920	T/ <u>C</u>	0.07	0.57 (0.14)	0.00003		
rs12535593	T/ <u>C</u>	0.27	-0.26 (0.07)	0.0002		

 Table 1: CD36 tag SNPs influence chylomicron levels.

Shown are beta-estimates ( $\beta$ ) <u>+</u> standard errors and raw p-values for the most significant *CD36* tag SNP (within respective LD blocks) that associate with chylomicron levels. SNPs are listed based on genomic positions (see Fig. 1). \*Bonferroni correction threshold p<0.006.

SNP	Effect allele	MAF	β(se)	* <i>p</i> -value			
	<u>0 - 3.5 h</u>						
rs304763	$G/\underline{A}$	0.38	-0.15 (0.05)	0.0049			
rs304798	A <u>/G</u>	0.21	-0.21 (0.07)	0.0019			
rs304802	T/ <u>C</u>	0.32	-0.20 (0.06)	0.0009			
<u>3.5 - 6.0 h</u>							
rs37920	T/ <u>C</u>	0.07	0.44 (0.13)	0.0008			
rs13228738	G/ <u>A</u>	0.38	0.16 (0.05)	0.0032			
rs1931694	G/ <u>C</u>	0.47	0.15 (0.05)	0.0048			
rs3211842	G/ <u>A</u>	0.42	0.17 (0.05)	0.0013			
Shown are beta-estimates $(\beta)$ + standard errors and raw p-values for							

Table 2: Effect of CD36 SNP on growth curve slopes for TGappearance (0 - 3.5 h) and clearance (3.5 - 6.0 h) after the meal.

Shown are beta-estimates ( $\beta$ ) + standard errors and raw p-values for the most significant CD36 tag SNP (within respective LD blocks). For TG appearance negative  $\beta$  estimates indicate the allele reduces absorption. For clearance, negative  $\beta$  indicates enhanced TG removal and positive  $\beta$  reduced TG removal.\*Bonferroni correction threshold p<0.006.

SNP	Effect Allele	MAF	β(se)	<i>p</i> -value
rs6970109	C/ <u>A</u>	0.10	0.26 (0.09)	0.0027
rs1722507 <sup>b</sup>	A/ <u>G</u>	0.39	0.17 (0.05)	0.0011
rs1761665 <sup>a,b</sup>	T/ <u>C</u>	0.44	0.20 (0.05)	0.0001
rs3211842ª	$G/\underline{A}$	0.42	0.02 (0.05)	0.0017
rs7755°	$G/\underline{A}$	0.46	0.16 (0.05)	0.0020

Table 3: CD36 tag SNPs associate with higher LDL.

SNPs are listed by genomic location. Effect allele is bolded. Shown are beta-estimates ( $\beta$ )  $\pm$  standard errors and raw p-values for most significant CD36 tag SNP (within respective LD blocks). \*Bonferroni correction threshold p<0.006. LD between SNPs: <sup>a</sup>r<sup>2</sup>>0.80, D'=0.93, <sup>b</sup>r<sup>2</sup>=0.68, D'=1.0, <sup>c</sup>r<sup>2</sup>>0.80, D'=0.96. Data shown are for fasting LDL, Supplementary Table S2 shows data for 3.5 and 6h post meal.

		Adipose Tissue		Hea	rt <sup>c</sup>
SNP	Effect Allele	β(se)	<i>p</i> -value	β(se)	<i>p</i> -value
rs1931694 (TGc)	G/ <u>C</u>	-0.17 (0.03)	1.24 x 10 <sup>-09</sup>		
rs1722507 <sup>b</sup> (TGc)	A/ <u>G</u>	-0.25 (0.03)	5.58 x 10 <sup>-17</sup>		
rs1761665 <sup>a</sup> (LDL)	T <u>/C</u>	-0.27 (0.03)	3.31 x 10 <sup>-22</sup>	-0.40 (0.08)	7.25x10 <sup>-7</sup>
rs13228738 <sup>b</sup> (TGc)	$G/\underline{A}$	-0.24 (0.03)	1.03 x 10 <sup>-16</sup>		
rs3211842 <sup>a</sup> (TGc)	$G/\underline{A}$	-0.28 (0.03)	2.27 x 10 <sup>-22</sup>		
rs7755 (LDL)	G/ <u>A</u>	-0.44 (0.03)	1.12 x 10 <sup>-50</sup>	-0.53 (0.08)	4.32x10 <sup>-11</sup>

 Table 4:
 Lipid-associated SNPs influence CD36 transcript level.

Correlations between alleles and CD36 mRNA were determined using simple regression. SNPs are listed by effect size. Frequency of the effect allele is  $\geq$ 40% for all listed SNPs. The lipid trait affected by the SNP is in brackets, TGc = TG clearance. SNPs in LD  ${}^{a}r^{2}$ >0.80, D'>0.8),  ${}^{b}(r^{2}$ =0.93, D'>0.8). SNPs are listed based on genomic location. CD36 mRNA levels (ILMN 1784863; Illumina HT-12 v3) in abdominal subcutaneous adipose tissue in the Multiple Tissues Human Expression Resource (MuTHER) were retrieved from GENEVAR database. SNP effects on heart CD36 expression (ILMN 1784863; Illumina HumanHT-12 v4) from a genome wide eQTL dataset in non-diseased tissue of 129 Western Europeans (45). --- indicates effects were not reported.

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CpG ID	rs13228738	rs3211842	rs1722507	rs1761665	rs7755
cg18508525 <sup>a</sup>	9.65 x 10 <sup>-06</sup>	3.83 x 10 <sup>-04</sup>	2.76 x 10 <sup>-07</sup>	3.20 x 10 <sup>-05</sup>	3.04 x 10 <sup>-04</sup>
cg25783969 <sup>a</sup>	1.24 x 10 <sup>-08</sup>		7.23 x 10 <sup>-11</sup>	6.72 x 10 <sup>-05</sup>	9.26 x 10 <sup>-05</sup>
cg21055948 <sup>a</sup>	1.57 x 10 <sup>-11</sup>	5.01 x 10 <sup>-06</sup>	2.37 x 10 <sup>-13</sup>	3.08 x 10 <sup>-07</sup>	6.14 x 10 <sup>-06</sup>
cg19096849	3.36 x 10 <sup>-09</sup>	5.87 x 10 <sup>-06</sup>	3.64 x 10 <sup>-11</sup>	1.40 x 10 <sup>-07</sup>	7.97 x 10 <sup>-07</sup>
cg26138637 <sup>a</sup>			8.38 x 10 <sup>-05</sup>		
cg06601993			6.75 x 10 <sup>-04</sup>	4.00 x 10 <sup>-04</sup>	

Table 5: Lipid- SNPs associate with DNA methylation in adipose tissue.

Adipose tissue CpG methylation levels determined in the MuTHER study using the Infinium Human Methylation 450 platform. CpGs listed based on chromosomal position. P values are for CpG-SNP associations. <sup>a</sup>CpG associates with adipose *CD36* mRNA (Table 6). Table 6: Hypomethylation of promoter CpGsites associates with adipose tissue CD36expression levels.

CpG ID	<i>p</i> -value
cg18508525	2.40 x 10 <sup>-05</sup>
cg25783969	2.03 x 10 <sup>-05</sup>
cg21055948	3.06 x 10 <sup>-05</sup>
cg26138637	6.66 x 10 <sup>-05</sup>
CD36 gene expression	(ILMN_1784863) was
quantified in adipocyte nu	clei from 776 individuals

using Illumina whole genome expression array (Human HT-12 v3).

-

CpG ID	TG		LDL	
	β	p-value	β	p-value
cg18508525 <sup>a</sup>	0.24	1.36 x 10 <sup>-06</sup>	0.17	3.05 x 10 <sup>-05</sup>
cg25783969 <sup>a</sup>	0.21	5.17 x 10 <sup>-04</sup>	0.11	1.49 x 10 <sup>-02</sup>
cg21055948 <sup>a</sup>	0.22	3.99 x 10 <sup>-04</sup>	0.12	7.23 x 10 <sup>-03</sup>
cg26138637 <sup>a</sup>	0.20	1.12 x 10 <sup>-05</sup>	0.13	$2.82 \times 10^{-03}$

Table 7: Methylation at CD36 expression-associated CpG sites correlateswith TG and LDL concentrations.

Adipose tissue CpG methylation levels were determined in the MuTHER study using the Infinium Human Methylation 450 platform. CpGs are listed based on chromosomal position. <sup>a</sup>CpG associates with adipose *CD36* mRNA (Table 6).

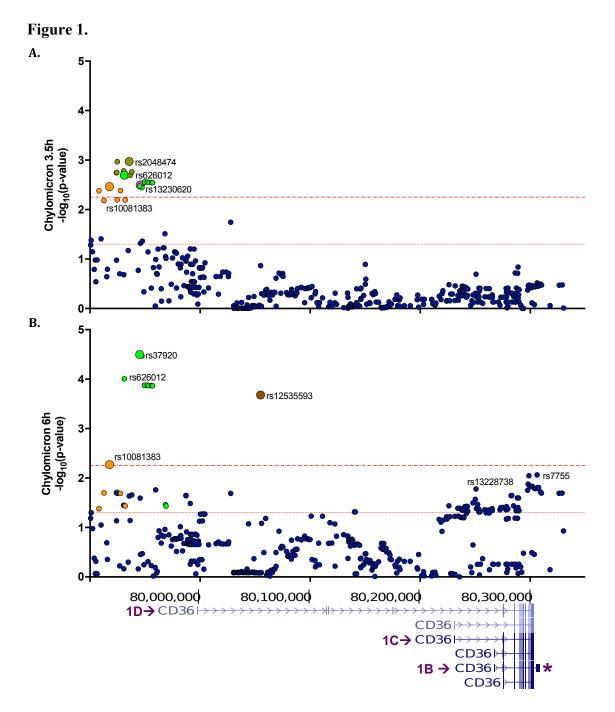


Figure 1: SNPs near distal first exon 1D of *CD36* associate with Chylomicron remnants. Regional plot of *CD36* SNPs that associate with chylomicron remnant concentration at A) 3.5h and B) 6 h after a the fat meal. P-values ( $-\log_{10}$ ) are plotted based on chromosomal position (human genome GRCh37/hg19). Colors differentiate LD blocks ( $r^2 \ge 0.8$ ). The most significant SNP (representative tag SNP) is identified by a larger symbol and the rs ID number. The overlay of the *CD36* gene from the UCSC genome browser illustrates the SNP position relative to alternative *CD36* promoters and transcripts (referred to as 1B, 1C, 1D) and the long 3'UTR (asterisk). Dotted red lines indicates p-value  $\le 0.05$  and the Bonferroni-corrected threshold, p<0.006.



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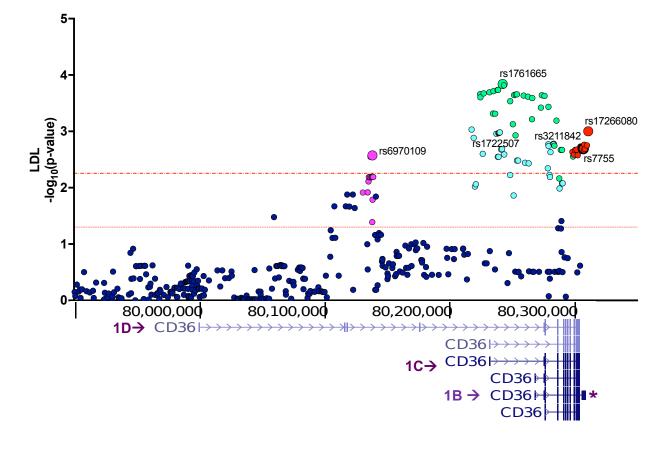


Figure 2: CD36 SNPs in the proximal and central promoter regions associate with fasting LDL. P-values (-log<sub>10</sub>) are plotted based on chromosomal position (human genome GRCh37/hg19). Colors differentiate LD blocks ( $r^2 \ge 0.8$ ) with the most significant SNP (representative tag) identified by a larger symbol and rs ID number. Dotted red lines indicates pvalue  $\leq 0.05$  and Bonferroni threshold, p<0.006.

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Figure 3.
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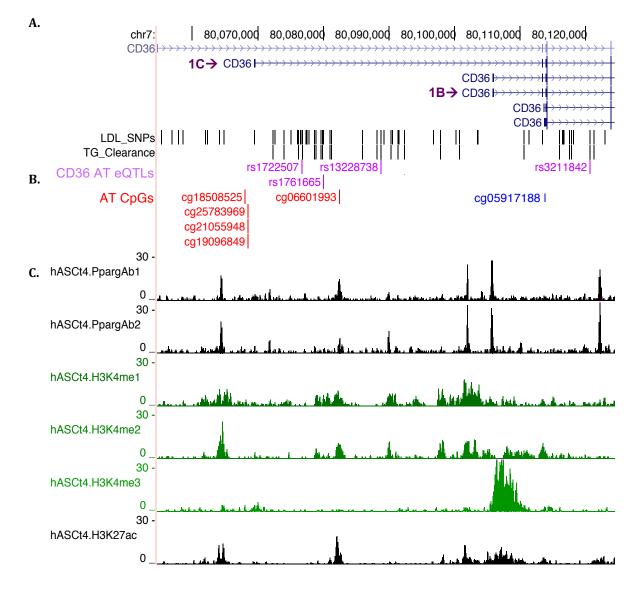


Figure 3: TG Clearance and LDL associated CD36 SNPs align in proximity to PPARy motifs and to Cpg sites in the CD36 promoter. The central CD36 promoter region (chr7: 80,060,000-80,095,000) is tagged by rs1761665 which associates with methylation levels at CpG sites in adipose tissue near CD36 alternative exon 1C that is specific to adipocytes (absent from immune cells and endothelial cells). (A) Purple text indicates SNPs that associate with both lipid levels and adipose tissue CD36 expression. (B) SNP-associated methylation sites (red text). The location of a previously published TG-associated CpG site is shown in blue text (47). (C) Chromatin state maps of the CD36 locus from mature human adipocytes. Histograms of ChIP-Seq fragments aligned to the human reference genome (hg18) highlighting active PPARy binding sites (peaks) and epigenetic signatures (H3K4me1, H3K4me2, H3K4me3, H3K27ac).

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