

# NIH Public Access

**Author Manuscript**

Am J Hum Biol. Author manuscript; available in PMC 2014 September 01.

## Published in final edited form as:

Am J Hum Biol. 2013 September ; 25(5): 673–680. doi:10.1002/ajhb.22429.

## **Evidence for novel genetic loci associated with metabolic traits in Yup'ik people**

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## **Abstract**

**Objectives:** To identify genomic regions associated with fasting plasma lipid profiles, insulin, glucose, and glycosylated hemoglobin in a Yup'ik study population, and to evaluate whether the observed associations between genetic factors and metabolic traits were modified by dietary intake of marine derived omega-3 polyunsaturated acids (n-3 PUFA).

**Methods:** A genome-wide linkage scan was conducted among 982 participants of the Center for Alaska Native Health Research study. n-3 PUFA intake was estimated using the nitrogen stable isotope ratio  $(^{15}N)$  of erythrocytes. All genotyped SNPs located within genomic regions with LOD scores > 2 were subsequently tested for individual SNP associations with metabolic traits using linear models that account for familial correlation as well as age, sex, community group and n-3 PUFA intake. Separate linear models were fit to evaluate interactions between the genotype of interest and n-3 PUFA intake.

**Results:** We identified several chromosomal regions linked to serum apolipoprotein A2, high density lipoprotein-, low density lipoprotein-, and total cholesterol, insulin, and glycosylated hemoglobin. Genetic variants found to be associated with total cholesterol mapped to a region containing previously validated lipid loci on chromosome 19, and additional novel peaks of biological interest were identified at 11q12.2-11q13.2. We did not observe any significant interactions between n-3 PUFA intake, genotypes, and metabolic traits.

**Conclusions:** We have completed a whole genome linkage scan for metabolic traits in Native Alaskans, confirming previously identified loci, and offering preliminary evidence of novel loci implicated in chronic disease pathogenesis in this population.

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## **Keywords**

Alaska Native; metabolism; multi-point linkage genome scan

The prevalence of type 2 diabetes (T2D) in Alaska Native communities has historically been lower than the general US population, although the specific genetic and environmental factors that underlie these statistics remain poorly characterized. Accumulating evidence suggests that elements of the traditional diet such as omega-3 polyunsaturated fatty acids  $(n-3$  PUFA) and low carbohydrate content (Naylor *et al.*, 2003) may contribute to the low T2D prevalence in these communities. More recently, Alaska Native people have experienced a more dramatic rise in T2D burden than other Native groups, with the ageadjusted prevalence increasing from  $17.3/1,000$  in 1985 to  $47.6/1,000$  (Narayanan *et al.*, 2010). This increase may be attributed to increased awareness, improved access to healthcare, decreased overall mortality rates, a change in diagnostic criteria in 1997, as well as nutritional and lifestyle transitions (Naylor et al., 2003). Moreover, the effects of the changing diet on chronic metabolic disease risk may be mediated by genetic factors that are unique to isolated populations (Lemas et al., 2013).

Current understanding of the genetics of metabolic traits among circumpolar populations is limited in comparison to other ethnic groups, highlighting the need for comprehensive scans to identify the major susceptibility loci. Two recent candidate gene studies, conducted among the Canadian Inuit (Rudkowska et al., 2013) and Yup'ik people living in Southwest Alaska (Lemas et al., 2012) suggested that consumption of marine-derived n-3 PUFA may influence lipid traits through interactions with common genetic variants implicated in metabolic pathways. However, few studies have comprehensively examined the role of genes or gene-diet interactions in the etiology of metabolic traits in communities with a historically low prevalence of T2D, despite ample evidence of their etiologic relevance in other populations (Ntzani and Kawoura, 2012; Franks et al., 2007; Cole et al., 2005).

Rudkowska et al. showed that the minor allele frequency distribution of several metabolismrelated genetic markers differs markedly between the Inuit and Caucasians, and other studies of complex traits (Hegele et al., 1997; Hegele et al., 1999) also suggest differences in susceptibility genes. To further elucidate these differences, the purpose of this study was to examine the genetic architecture of metabolic traits in a Yup'ik population using wholegenome linkage analysis, followed by targeted association testing for variants genotyped within the observed peaks. Additionally, this study investigated potential interactions of these loci with habitual intake of marine-derived n-3 PUFA.

## **MATERIALS AND METHODS**

#### **Participants**

The Center for Alaska Native Health Research (CANHR) studies genetic, behavioral, and dietary risk factors underlying obesity and their relationship to diabetes and cardiovascular disease among the Yup'ik people (Mohatt *et al.*, 2007). Recruitment of Yup'ik participants was initiated in 2003 and continues in 11 Southwest Alaska communities. All residents are invited to participate, with the study sample age distribution reflecting that of all eligible participants as summarized by the 2000 U.S. census. The analyses in this report were performed on 1136 non-pregnant Yup'ik people that were 14 years old at the time of enrollment.

#### **Ethics**

Participants provided written informed consent using protocols approved by the University of Alaska Institutional Review Board, the National and Alaska Area Indian Health Service Institutional Review Boards, and the Yukon Kuskokwim Human Studies Committee.

#### **Measurements of metabolic traits**

A radioimmunoassay kit was used to assay fasting insulin (FI) with an I125-iodinated insulin tracer, anti-human insulin specific antibody, and human insulin standards (Linco Research, St Charles, MO, USA), for which the intra- and inter-assay variations were respectively 5.8% and 10.2%. A Cholestech LDX analyzer was used to measure fasting glucose (FG), and glycosylated hemoglobin (HbA1c) was quantified with a DCA 2000+ analyzer (Bayer AG, Leverkusen, Germany). Lipid profiles, specifically low-density lipoprotein (LDL) cholesterol, high-density lipoprotein (HDL) cholesterol, very low-density lipoprotein (VLDL) cholesterol, total cholesterol, triglycerides (TG), and apolipoproteins A1 (ApoA1) and A2 (ApoA2) were measured as described in previous publications from our group (Lemas *et al.*, 2012; Boyer *et al.*, 2007).

## **Dietary n-3 PUFA intake measurements**

The nitrogen stable isotope ratio ( $^{15}N$ ) of red blood cells (RBC) was used to assess n-3 PUFA intake as previously described (O'Brien et al., 2009). RBC aliquots were autoclaved for 20 minutes at 121° C to destroy blood-borne pathogens, and samples were weighed into 3.5 × 3.75 mm tin capsules and freeze dried to a final mass of 0.2 - 0.4 mg. Samples were analyzed at the Alaska Stable Isotope Facility by continuous-flow isotope ratio mass spectrometry, using a Costech ECS4010 Elemental Analyzer (Costech Analytical Technologies, Valencia, CA, USA) interfaced to a Finnigan Delta Plus XP isotope ratio mass spectrometer via the Conflo III interface (Thermo-Finnigan Inc., Bremen, Germany). The conventional method of expressing nitrogen isotope ratios at natural abundance is in "permil" (%<sub>0</sub>) abundance of <sup>15</sup>N relative to atmospheric nitrogen (<sup>15</sup>N/<sup>14</sup>N<sub>atm</sub> = 0.0036765), as follows:

 $^{15}N = [(15N/14N_{sample} - 15N/14N_{atm})/(15N/14N_{atm})] \cdot 1000\%$  D. Laboratory reference materials (peptone,  $15N = 7.00$ ) were concurrently prepared and run to assess analytical accuracy and precision; these were analyzed after every eighth sample and estimated  $15N$  at  $7.0 \pm 0.2\%$  (mean  $\pm$  SD). The range of isotopic variation in the data (9%<sub>0</sub>) was very large relative to analytical precision  $(0.2\%)$ . To avoid undue influence of extreme values and allow for a non-linear relationship suggested by prior evidence, this variable was modeled as a categorical variable, defined by the quartiles of the  $15N$  measurements. We identified the quartiles of values of  $15N$ , and assigned values to three dichotomous variables as follows. Individuals with values below the first quartile were assigned  $(0,0,0)$  to the three variables. Those with values at least equal to the first quartile but below the second were given the values (1,0,0). Those with values at least equal to the second quartile but below the third were given the values  $(1,1,0)$ . Finally, those with values at least equal to the third quartile were given the values  $(1,1,1)$ .

## **Genotyping**

Participants were genotyped for 6090 single nucleotide polymorphisms (SNPs) from the Illumina's Linkage-IV panel (Illumina, San Diego, CA, USA) at the Center for Inherited Disease Research (CIDR). The Illumina Linkage-12 panel spans the entire genome with an average genetic distance of 0.58 cM and 441 kb physical map spacing.

## **Pedigree analysis**

Pedigree data was extracted from the Progeny database (Progeny Software LLC, South Bend, IN, USA), and merged as appropriate into larger pedigrees using the program Pedmerge Pedmerge allows for accurate and efficient merger of separately ascertained pedigrees that belong to the same extended family (Plaetke and Balbi, 2010). The 1136 individuals included in this study were members of 63 independent pedigrees. Among these, 21 pedigrees contained at least two individuals and 42 individuals were singletons; both categories were subsequently excluded from the linkage analyses. The majority of the genotyped participants  $(n=1037)$  were members of the single largest pedigree, which was too large to analyze with standard frequentist methods. For this reason, that pedigree was cut using Pedstr into 113 smaller, independent sub-pedigrees each containing between 2 and 22 genotyped participants. Pedstr is an automatic algorithm that splits large pedigrees into fragments of specified size, which can subsequently be used in multi-point linkage analysis (Kirichenko et al., 2009). Of the total 938 genotyped participants in these cut pedigrees, 52 were repeated once or twice (e.g. as a founder in one pedigree and a child in another) to maintain close relationships; 883 of the original genotyped participants were retained. The remaining 154 genotyped individuals were not closely related to any of the retained participants. A summary of the relative pairs included in the analysis is given in Table 1.

## **Quality control of phenotypic and genotypic data**

Simple linear models were fitted to outcome variables using all available non-genetic covariates, and descriptive statistics of the residuals were examined using R software (v2.10.1, R Development Core, 2009). If necessary, traits were transformed using either logarithmic (for HDL and TG) or Box-Cox approaches (Box and Cox, 1964) (for ApoA1, ApoA2, FG, FI, HbA1C) to reduce the residual kurtosis of the traits and avoid artificially inflated evidence for linkage (Blangero et al., 2001). Estimates of heritability and residual kurtosis for raw or appropriately transformed traits are given in Table 2. All traits exhibited statistically significant heritability estimates. For the traits that still had residual kurtosis above 0.8 (ApoA2, FG, HbA1c, FI, and untransformed LDL), subsequent linkage analyses used the t-distribution option to control type I error.

The genotypic data was subject to several quality control measures prior to whole-genome linkage analysis. First, Pedcheck was used to exclude SNPs that were not consistent with Mendelian segregation. Pedcheck is a program designed to check pedigree files to identify genotype incompatibilities in linkage analysis (O'Connell and Weeks, 1998). Second, SNPs that deviated from Hardy-Weinberg equilibrium (HWE) proportions (p<0.0001) were identified using an algorithm that considers familial relationships (Bourgain *et al.*, 2004) and excluded from the analysis. Finally, the program Hclust was used to select a subset of SNPs with low pairwise linkage disequilibrium (LD) to further reduce inflation of linkage evidence. Hclust performs hierarchical cluster analysis to identify the most parsimonious set of SNPs for further analysis based on linkage disequilibrium (Rinaldo et al., 2005)

All genotypes that passed the above quality checks, including those excluded by the Hclust algorithm, were used to ascertain cryptic population substructure using the PCA (Principal Components of Analysis) program from the Eigenstrat package. Eigenstrat was developed to detect and correct for population stratification in genome-wide association studies (Price et al., 2006). The computed values of the each component were grouped according to the villages of residence. Although the first component had no obvious systematic structure, the second component naturally split the genotyped sample into two groups, consistent with proximity of the villages to the coast. This observation was used to define "community group" as a dichotomous covariate included in subsequent analyses. The distribution of the second principal component of ancestry is shown as a box plot in Figure 1. Communities 1,

2, 3, 8, 9, and 11 (group 1) are located closer to the coast, while the remaining communities (group 2) are considered upriver (tundra) communities.

## **Linkage analysis**

SOLAR (Sequential Oligogenic Linkage Analysis Routines) was used to estimate the heritability and to test for whole-genome linkage. SOLAR is an extensive software package for genetic variance components analysis, including linkage (Almasy and Blangero, 1998). The variance component approach implemented in SOLAR requires an estimate of locus specific allelic sharing identical by descent (IBD), which was computed using MERLIN (Multipoint Engine for Rapid Likelihood Inference). MERLIN uses the Lander-Green approach to perform multipoint linkage analysis(Abecasis et al., 2002). All models run in SOLAR included age, sex, community group, and n-3 PUFA intake. In order to allow for non-linear effects of continuous covariates (age and n-3 PUFA intake) without undue influence of extreme values, both of these covariates were included in the models as three dichotomous variables to indicate quartiles.

#### **Association analysis**

All genotyped SNPs under linkage peaks identified in the linkage analysis that exceeded a LOD score of 2 and satisfied HWE constraints were tested for association with the traits using genetic association program ASSOC based on methods developed in George and Elston (1987) and Elston et al. (1992), which models the full familial correlation structure of the pedigree. ASSOC is a part of the Statistical Analysis for Genetic Epidemiology (S.A.G.E) suite of software, and is used in pedigree studies to estimate associations between traits and covariates as well as familial correlations, or heritability. The ASSOC program is able to handle arbitrary pedigrees, so the original pedigree was used rather than the pedigree processed by Pedstr (Kirichenko et al., 2009). All association models fit using ASSOC included the same non-genetic covariates that were used for the linkage analysis: sex, age, community group, and  $15N$  quartiles. For each SNP, three models were run: a baseline model containing only non-genetic covariates; a model containing non-genetic covariates and the SNP genotype as an additive effect; and a model containing non-genetic covariates, SNP genotype as an additive effect, and interaction between SNP effect and <sup>15</sup>N quartiles. Likelihood ratio tests were used to compare the models.

To correct for multiple testing in the association analyses, the effective number of SNPs was determined using spectral decomposition of the correlation matrix (Li and Ji, 2005; Nyholt, 2004).

## **RESULTS**

Clinical and demographic characteristics of the study population are presented by gender in Table 3. Of 982 individuals in the study, approximately 3% (n=29) had diagnosed T2D. Women represented slightly more than half of the overall study population, although that proportion was higher (57%) in the upriver communities. Consistent with a recent report, Yup'ik women had a higher intake of n-3 PUFAs as ascertained by  $15N$  (Nash *et al.*, 2012), as well as higher average fasting glucose, HbA1c and insulin relative to Yup'ik men. While circulating triglycerides, LDL, and VLDL cholesterol did not vary by gender, Yup'ik women had higher levels of total and HDL cholesterol as well as ApoA1, and Yup'ik men had higher levels of ApoA2.

All linkage findings that exceeded the cutoff LOD score of 2 are presented as genome-wide plots in Figure 2 and the corresponding linkage peaks are summarized in Table 4. In general, our whole genome linkage analyses showed stronger evidence for genetic linkage with

serum lipids relative to diabetes-related traits (FI, FG, or HbA1C). Specifically, we found a region on chromosome 11 (11p13-11q13.1) with the most robust evidence for linkage with LDL cholesterol (LOD score=3.02) that overlapped with the linkage peak for total cholesterol (LOD score= 2.00). Additionally, we observed a region of interest on chromosome 10 (10q23.1-10q24.1) which contained linkage peaks for HDL cholesterol (LOD score= 2.72) and triglycerides (LOD= 2.21). We did not detect chromosomal regions linked with ApoA1 or VLDL cholesterol.

Table 5 presents the results of genetic association testing for individual SNPs located in regions identified by the genome-wide linkage scan. We identified four SNPs that were nominally (P<0.05) associated with HDL cholesterol (P-values ranging from 0.0003 to 0.002), two that were associated with total cholesterol (P-values <0.006), and one associated with ApoA2 ( $P=0.0009$ ). However, the two markers on chromosome 11 that were significant hits for total cholesterol were determined to be in perfect LD  $(r^2=1)$  by the SNAP algorithm (Broad Institute, Cambridge, MA). As shown in Table 5, we did not detect SNPs associations with metabolic traits that were modified by n-3 PUFA intake.

## **DISCUSSION**

In this study, we conducted a whole-genome linkage analysis in a Yup'ik population with a historically low prevalence of T2D and identified chromosomal regions associated with meatbolic traits. In addition to observing genetic linkage to several known loci associated with plasma lipid/lipoprotein parameters on chromosome 19, these data demonstrate preliminary evidence of novel candidate genetic determinants of several lipid traits. Of particular biological interest is the high LD region on chromosome 11 containing MACROD1 and NAA40, which was found to be associated with measurements of total cholesterol in this sample of Yup'ik people.

MACROD1, also known as LRP16, is a gene with evidence of pleiotropic metabolic effects. A recent large-scale meta-analysis of genome-wide association studies (Winkler et al., 2012) has identified a locus in that gene as a sex-specific determinant of obesity-related traits like the waist-to-hip ratio, showing a robust association in women but not men. Given that women in Arctic communities have some of the highest visceral obesity rates in the world (Risica et al., 2000; Voruganti et al., 2006), MACROD1/LRP16 variation in this population may be an important contributor to health outcomes. Additionally, MACROD1/LRP16 has been shown to activate nuclear factor B (Wu et al., 2011), a key inflammatory component of the pathogenesis of atherosclerosis (Kutuk and Basaga, 2003) that can be influenced by environmental factors such as dietary cholesterol or statin therapy (Jasinska et al., 2007). It is plausible that mutations in MACROD1/LRP1 that result in the decreased expression of these genes have protective cardiometabolic effects in this population.

The quantitative trait loci (QTL) for HDL cholesterol on chromosome 19 are important for the interpretation of the study results. First, they are confirmatory linkage results for this study, as the identified genomic regions contain several loci—located in CILP2 for HDL cholesterol and *APOE/FLJ36070* for total cholesterol—that have been shown to be robustly associated with lipid levels in a variety of populations (Teslovich et al., 2010; Yan et al., 2011; Zhou et al., 2011). Additionally, they are consistent with data from another coastal Alaska Native study population, the Genetics of Coronary Artery Disease in Alaska Natives (GOCADAN) study (Cole et al., 2005), which identified a QTL for HDL levels (LOD=3.9) and suggestive evidence for linkage of LDL size (LOD=2.5) and HDL size (LOD=2.3) to the same genetic region. Cole *et al.* postulate that the LDL receptor gene  $(LDLR)$  represents a strong positional candidate for this QTL, as it is located 1.5 Mbp from the peak evidence of linkage. Second, the observed peaks are adjacent to the region containing QTL for several

obesity-related anthropometric traits found the GOCADAN study (Voruganti et al., 2011). This observation is consistent with another set of data from the GOCADAN population (Voruganti et al., 2006) which suggests that cholesterol phenotypes and obesity-related factors have shared genetic determinants in Alaska Native people. Future studies of putative genetic associations with these phenotypes should account for such pleiotropic effects to improve estimate validity and precision (Park et al., 2011).

Dietary n-3 PUFA intake did not modify the effect of genetic polymorphisms in the regions of interest on phenotypes in this Yup'ik study population. Despite the biological plausibility of such interactions, the null results of the interaction models do not support the role of longterm PUFA in modifying the observed genotype-phenotype associations. However, a recent study from the same population established that n-3 PUFAs may mediate the effect of obesity-related SNPs on adiposity (Lemas et al., 2013).

These findings must be interpreted in light of several important considerations. First, metabolic traits are complex and heterogeneous phenotypes that are likely to be influenced by a combination of both rare and common genetic variants and multiple dietary and lifestyle effects. While linkage analysis can efficiently detect families enriched with rare biologically relevant polymorphisms, this method may not have sufficient power to detect common variants of small or moderate effect size (Risch and Merikangas, 1996). To address that limitation, we supplemented our whole genome linkage analysis by testing individual SNPs within the linkage panel for association with lipid and diabetes-related traits. Notably, several linkage peaks identified in our study were concordant with results from genomewide association studies in other populations (Teslovich *et al.*, 2010). Second, the genetic regions identified in this study represent preliminary evidence, which should be independently replicated and followed with with functional genomic analyses in other populations with low T2D prevalence. Finally, the specific SNPs that were found to be significantly associated with metabolic traits may not represent the true susceptibility variant but may be in LD with the functional polymorphism, precluding any causal interpretation of these findings.

In conclusion, this study presents preliminary evidence of linkage between several genomic regions and metabolic traits in a Yup'ik population with a historically low prevalence of T2D. These results support the previous findings of lipid loci on chromosome 19 and highlight the potential importance of *MACROD1* to cardiometabolic health. Upon successful validation, these findings may serve to elucidate the heritable determinants of chronic disease in these genetically isolated, understudied communities and inform future research and clinical efforts.

## **Acknowledgments**

We thank the community field research assistants for helping with the study recruitment and data collection, and all study participants and their communities for welcoming and teaching our research team so much about the Yup'ik way of life. Genotyping services were provided by the Center for Inherited Disease Research (CIDR). CIDR is fully funded through a federal contract from the National Institutes of Health to The Johns Hopkins University, contract No. HHSN268200782096C.

**Funding**: NIH DK074842, RR016430, and K01DK080188

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## **Linkage Results for ApoA2**



## **Linkage Results for Insulin**









**Linkage Results for HDL** 





## **Linkage Results for Lab Triglycerides**



## **Linkage Results for LDL**



#### **Figure 2.**

Genome-wide linkage scans for selected metabolic traits in Yup'ik people. The X-axis shows the chromosomal location, and the Y-axis displays the LOD score.

- A) Apolipoprotein A2
- B) Fasting insulin
- C) Glycosylated hemoglobin
- D) High-density lipoprotein cholesterol
- E) Total cholesterol
- F) Triglycerides
- G) Low-density lipoprotein cholesterol

Number of pairs of study participants by degree of relationship.



Heritability estimates of metabolic trait measurements.<sup>1</sup>



1 ApoA1, apolipoprotein A1; ApoA2, apolipoprotein A2; HbA1C, glycosylated hemoglobin; HDL, high density lipoprotein; LDL, low density lipoprotein; VLDL, very low density lipoprotein

Descriptive characteristics of the study population (n=982).



 $1$  Mean (95% confidence interval) for continuous variables or n (% total) for categorical variables

 $P_{\text{Pcalculations are based on the 2-sample t-test, not accounting for the family structure.}$ 

 $\beta$  P calculations are based on the binomial distribution, not accounting for the family structure.

Linkage peaks exceeding LOD score of 2.



 $<sup>I</sup>$  Position was determined using hg18 as the reference genome.</sup>

Significant associations between single nucleotide polymorphisms within linkage peaks and metabolic traits in the study population.



 $^I\!$  Model included sex, age, n-3 PUFA intake, community group, and additive genetic term.

 $2$  Model included all terms listed above plus an interaction between the additive genotype and n-3 PUFA intake.

 $\beta$  P-value from the likelihood ratio test for interaction between the additive genotype and n-3 PUFA intake.