Characterization of the interindividual variability of lutein and zeaxanthin concentrations in the adipose tissue of healthy males and identification of combinations of genetic variants associated with it

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Supporting Tables, Figures and Methods

Supplemental Table S1. Candidate genes selected.

Genes that have been associated with serum XANTH concentration in genome wide association studies

Genes whose SNPs have been associated with the postprandial chylomicron carotenoid or triacylglycerol response in the same group of participants

Supplemental Figure S1. Candidate SNPs selection flowchart.

^aDeviations from Hardy-Weinberg equilibrium (HWE) can indicate inbreeding, population stratification, and genotyping errors.

Supplemental Table S3A-B. Comparison between adipose tissue lutein (L) and zeaxanthin (Z) concentrations measured at fast and 8 h after consumption of the 3 test meals.

A. Linear mixed model

^aUnstructured linear mixed model. Adipose tissue L and Z concentrations measured at fast and 8 h after consumption of the 3 test meals were analyzed with linear mixed models, using a full factorial design with meal (control, vitamin E and tomato puree) and time (fasting and 8 h postmeal) as fixed within-subject variables and participant as the random variable. Of the 5 linear mixed models tested, the unstructured model was selected based on Akaike's Information Criterion ²³

B. Paired *t*-test

Abbr: df, degrees of freedom ; sig, significance (*p*-value) for paired *t*-test

Supplemental Table S4. Characteristics of the partial least squares regression models ^a

A. Lutein

^aThe selected model is highlighted in bold font. All models had 1 component.

^bSee Equation 1.

^cSee **Supplemental Figure S3** for further explanation of the procedure.

 d See 24 .

Pairwise linkage disequilibrium (LD) test: Identification of LD SNPs in the final PLS model

The PLS regression model chosen for zeaxanthin contained 13 SNPs (**Table 3B**, main manuscript). Among these, 3 were in LD according to an online calculator tool for pairwise LD (available at https://www.ensembl.org/Homo_sapiens/Tools/LD, population: European CEU, queried in 5 April 2024). Since these SNPs conveyed redundant information in the model, we kept the one which had the higher VIP value, leaving 11 SNPs in the final selected PLS regression model (Table 3B).

Supplemental Table S5. SNPs in LD in the final PLS regression model.^a

^a Gene names can be found in Supplemental Table S1.

^b VIP: Variable Importance in the Projection

^cPairwise LD *R*² results were generated from population *1000GENOMES:phase_3:CEU (accessible at* https://www.ensembl.org/Homo_sapiens/Tools/LD)

Supplemental Information: additional validations of the partial least squares (PLS) regression model.

*1) Leave-*k*-out* cross-validation

The leave-*k*-out validation was based on Steyerberg *et al.*²⁵ . Briefly, the two selected PLS regression models were accessed by randomly taking out *k* participants $(k=1,2,3,4)$ from the original dataset, thus leaving a training dataset. The *k* participants taken out were then reintroduced into this training set to assess whether the models built without these *k* participants would be able to predict accurately their adipose tissue Xanth concentrations. This test was performed up to 42 times so that each participant was taken out once (*i.e.* 42 times for *k*=1, 21 times for $k=2$, 14 times for $k=3$ and 10 times for $k=4$).

The relative prediction error (%) between the predicted and the measured adipose tissue Xanth concentrations for each *k* are presented below (**Supplemental Table S6)**. The percentage of error remained relatively stable, after leave-*k*-out cross-validation test, suggesting that the PLS regression model was relatively robust.

Supplemental Table S6. Average relative prediction error (%) following the leave-*k*-out procedure.

*2) Regression coefficient stability testing following the leave-*k*-out procedure*

The stability of the regression coefficients of each selected model (**Table 4**) remained unchanged (*p*>0.05; ANOVA) even when up to 4 participants were left out of the model (*detailed procedure is shown in Supplemental information 1*). **Supplemental Figure S2-A (lutein) and**

S2-B (zeaxanthin) shows good stability of the regression coefficients with this validation.

Regression coefficient stability test: Lutein

Supplemental Figure S2-A: Regression coefficient stability validation for adipose tissue lutein concentration following the leave-*k*-out procedure. *k* participants (*k*={1,2,3,4}) were randomly removed from the original dataset, thus leaving a training subset. These participants were

then reintroduced in the training subset to assess the regression coefficients of the 7 SNPs from the selected model. This test was performed at most 42 times so that each participant was taken out once. One-way ANOVA performed for each gene showed no significant differences between the full model and the 4 training subsets generated by the procedure. Gene names are found in **Supplemental Table 1**.

Supplemental Figure S2-B: Regression coefficient stability validation for adipose tissue zeaxanthin concentration following the leave-*k*-out procedure. *k* participants (*k*={1,2,3,4}) were randomly removed from the original dataset, thus leaving a training subset. These participants were then reintroduced in the training subset to assess the regression coefficients of the 11 SNPs from the selected model. This test was performed at

most 42 times so that each participant was taken out once. One-way ANOVA performed for each gene showed no significant differences between the full model and the 4 training subsets generated by the procedure. Gene names are found in **Supplemental Table 1**.

3) R*² and adjusted* R*² of the selected model after 100 permutations.*

This procedure 1) assessed the risk that the PLS regression model is spurious, *i.e.* the model fits the current data set well but does not predict Y well for new observations, and 2) tested for over-fitting. Briefly, the accuracy of fit (*R*² and *R*² after cross-validation) of the original model was compared with the accuracy of fit of 100 models based on data where the order of the *Y* matrices for the participants (adipose tissue L and Z concentrations) were randomly permuted, while the *X* matrices (list of selected SNPs) were kept intact. A robust model (where the fit between *X* and *Y* is high) should not be able to correctly predict the permuted *Y* variables with the intact *X* variables. **Supplemental Figure S3-A (L) and supplemental Figure S3-B (Z)** present results of these permutations using their corresponding PLS regression model.

Supplemental Figure S3-A. The horizontal axis represents the correlation between the permuted *Y*'s and the original *Y*'s. The vertical axis represents the *R*² (dashed line and black triangles) and *R*² after cross-validation (dashed line and squares) values obtained in the permuted models. Values of the original model are on the far right (at correlation $= 1$) while values of the 100 *Y*-permuted models are further to the left. The average *R*² after 100 permutations was 0.11. This strongly supports the conclusion that the ability of the original, non-permuted model, to predict **Permutation test: Zeaxanthin**

Supplemental Figure S3-B. The horizontal axis represents the correlation between the permuted *Y*'s and the original *Y*'s. The vertical axis represents the *R*² (dashed line and black triangles) and *R*² after cross-validation (dashed line and squares) values obtained in the permuted models. Values of the original model are on the far right (at correlation $= 1$) while values of the 100 *Y*-permuted models are further to the left. The average $R²$ after 100

permutations was 0.18. This strongly supports the conclusion that the ability of the original, non-permuted model, to predict the adipose tissue zeaxanthin concentration is not due to chance.

(a) Adipose tissue lutein concentration

(b)Adipose tissue zeaxanthin concentration

Supplemental Figures S4 A and B. Retrospective multivariate power calculation was performed for 2 PLS models that incorporated the SNP variables associated with both adipose tissue lutein (7 SNPs) and zeaxanthin (11 SNPs) concentrations (refer to **Table 3**, main manuscript). With an FDR-adjusted p-value of 0.001, the predicted multivariate power of 80 % (on the vertical axis) was computed for a sample size of 15 per group for (a) adipose tissue lutein concentration and 20 per group for (b) adipose tissue zeaxanthin concentration (on the horizontal axis), affirming that the chosen sample sizes for this study (total sample size of 42 participants) were adequate. Image obtained from MetaboAnalyst 6.0 website, accessible at https://new.metaboanalyst.ca .

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