

Influence of FUT2/FUT3 polymorphism on human breastmilk oligosaccharide composition in the EDEN mother-child cohort

Eric Venot, Sylvain Dechaumet, Blanche Guillon, Mikaïl Berdi, Sophie Cholet, Ghita Tajeddine, Manon Mutaner, Aline Meirhaeghe, Barbara Heude, Blandine De Lauzon-Guillain, et al.

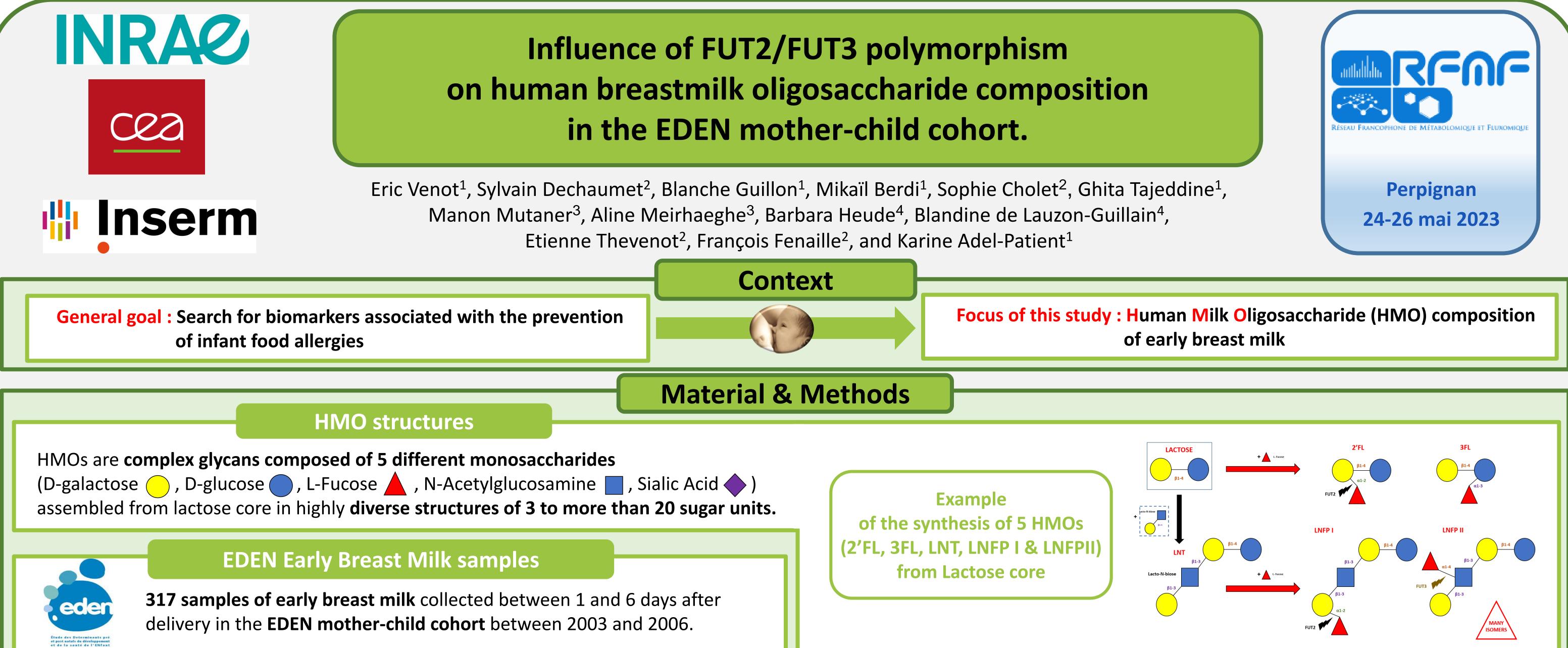
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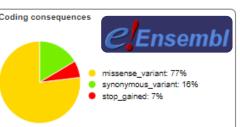
Secretor/Lewis genotypes

Maternal genetic predispositions induce HMO synthesis differences (Lefebvre et al., 2020): 2 fucosyltransferase enzymes production depends on mother FUT2 (Secretor) and FUT3 (Lewis) genes : - active FUT2 (Secretor + \checkmark) \Leftrightarrow 2'FL & LNFP I synthesis

- active FUT3 (Lewis + 1) \Leftrightarrow LNFP II synthesis

220 out of the 317 EDEN cohort mothers were genotyped : 62 SNPs associated to FUT2 and FUT3 genes were extracted. After SNP quality filters, selection of non homozygous SNPs for all EDEN mothers and selection of representative SNPs in complete Linkage Desequilibrium with others, **17 SNPs** were kept.

Ensembl Variant Effect Predictor tool was also applied on the 62 SNPs.



LC-HRMS analysis & HMO workflow

- The samples were analyzed by Liquid Chromatography coupled to High Resolution **Mass Spectrometry** (**Q-TOF**TM, res: 40000, PGC Hypercarb column)
- HMO references:
 - 19 commercial standards analyzed
 - m/z ratios of 8 953 HMO computed in silico based on
 - all possible monosaccharide combinations.
- specific in house HMO workflow from mzML files developed with to detect m/z ratios corresponding to potential HMOs





→ 130 HMOs detected after selection criteria including 18 HMOs annotated based on commercial standard (including **2'FL**, **LNFP I & LNFP II**)

Statistical analysis

Results

• The significance of the difference in intensity between genotype groups was assessed by univariate hypothesis testing with multiple testing correction FDR (Benjamini et al., 1995), (phenomis R package (Thévenot et al., 2023)), with : - Student t test when minor allele-minor allele genotype was compared to the 2 others gathered. - Analysis of variance when the 3 different genotypes were considered.

Hierarchical cluster analysis (HCA) (ComplexHeatmap R package (Gu et al., 2016)) was performed on all samples and all (annotated and potential) HMOs.

FUT2/FUT3 effects on HMO intensities

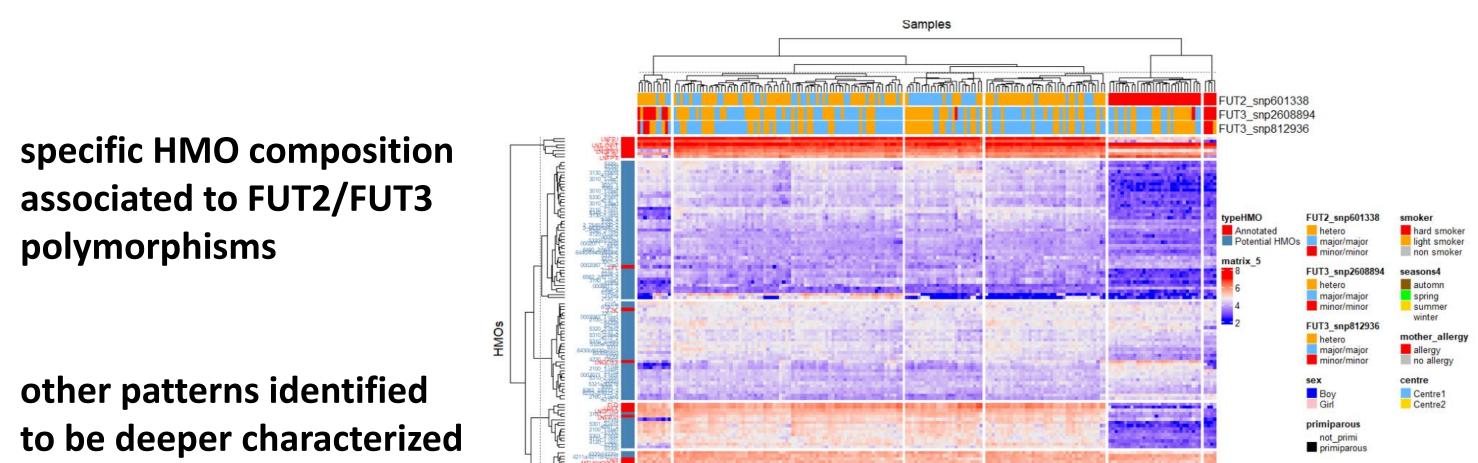
New impacted HMOs identified

- HMO composition highly depending of FUT2 and FUT3 polymorphisms :
 - FUT2 has an impact on 78% of the 130 detected HMOs (p-values limit : 0.001)
 - FUT3 effect is also important but weaker than FUT2 ones (27% of the HMOs **impacted** with 0.05 p-value limit)

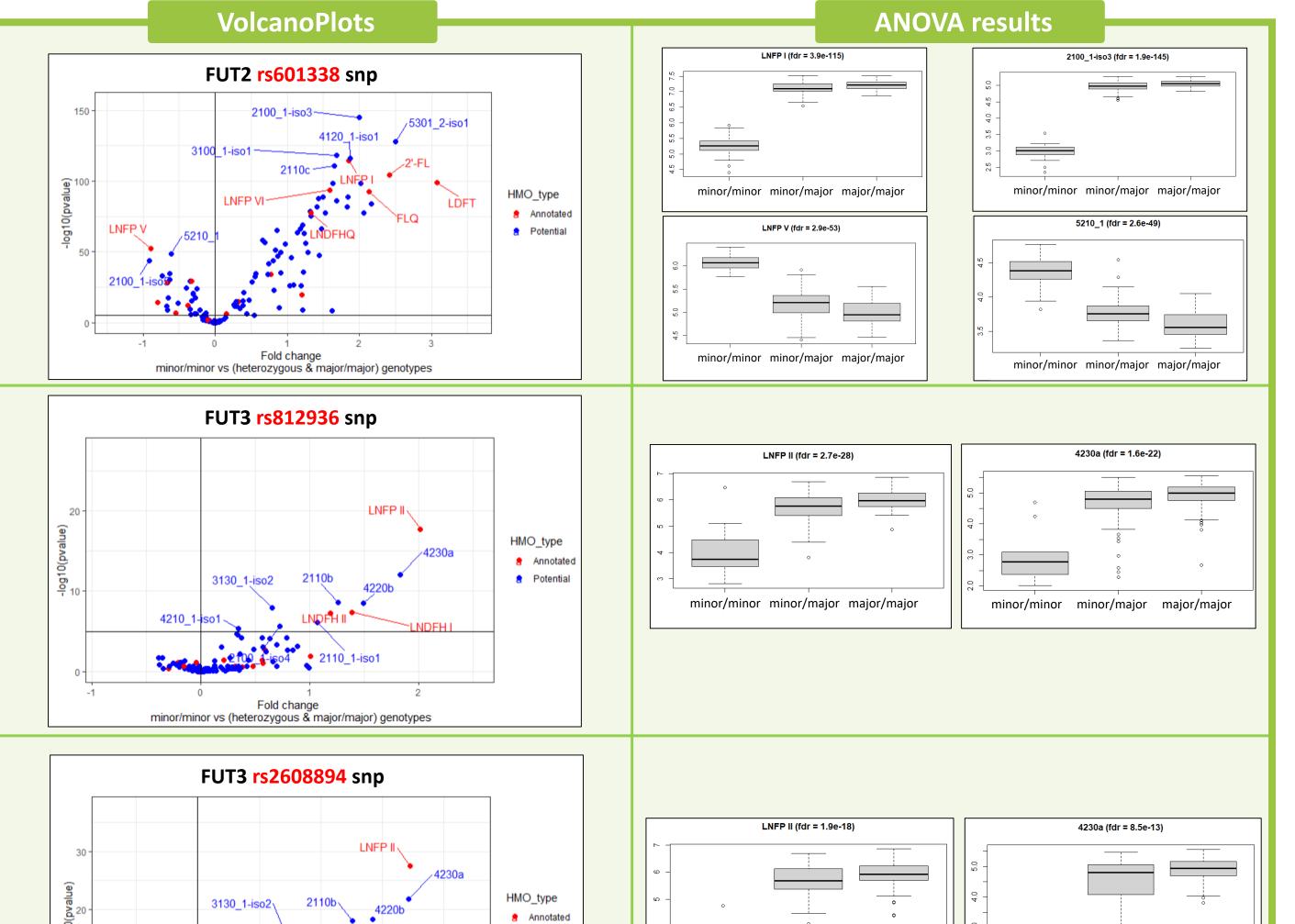
	snp	Ensembl Variant Effect Predictor type	Genotype distribution in 220 EDEN mothers			Considering the 3 genotypes separated				Considering the minor/minor genotype versus 2 other genotypes						
gene			minor /minor	minor /major	major /major	number of HMOs	pvalues			number of	pvalues			log10 difference		
							median	min	max	HMOs	median	min	max	median	min	max
FUT2	rs601338*	stop_gained	46	114	60	101	1.23e-34	1.91e-145	7.06e-05	98	2.27e-26	1.76E-54	7.04e-05	0.78	-1	3.06
FUT3	rs812936*	missense variant	130	82	8	35	7.86e-04	1.95e-18	4.81e-02	24	6.60e-03	3.20e-10	4.39e-02	0.59	-0.44	2.2
FUT3	rs778986	missense variant	134	79	7	35	5.75e-04	2.07e-18	4.20e-02	24	6.60e-03	3.20e-10	4.39e-02	0.62	-0.44	2.2
FUT3	rs2608894	regulatory region variant - promoter	108	96	16	33	1.46e-05	2.71e-28	3.22e-02	29	1.51e-04	4.54e-12	4.82e-02	0.51	-0.2	1.88

• **2** SNPs (* in the table) mentioned in Lefebvre et al. paper (2020) confirmed in this study. • 2 new SNPs with high impact detected in this study (rs778986 & rs2608894)

HCA results & different HMOs patterns identified



• Association between FUT2/FUT3 polymorphisms and several HMOs (2'FL, LNFP I, LNFP II...) already described in the literature (Lefebvre et al., 2020) found back in this study. • But new potential HMOs with higher or similar association levels with FUT2/FUT3 polymorphisms were also identified.



	sex primiparous smoker seasons4 mother_allergy centre	Image: State of the state						
 High impact of FUT2 & FUT3 polymorphisms on HMOs composition Results for 2 Lefebvre et al. SNPs confirmed and larger list of impacted HMOs identified Different clusters of HMOs identified 	 What's next? Deeper insight into different HMOs clusters (HMOs mother profiling?) Assessment of influence of environmental factors on milk HMO composition & association between HMO composition and health outcomes in childhood 	 Lefebvre et al., 2020. Time of lactation and maternal Fucosyltransferase genetic polymorphisms determine the variability of Human milk oligosaccharides. <i>Frontiers in Nutrition</i>. Smith et al., 2006. XCMS: Processing mass spectrometry data for metabolite profiling using nonlinear peak alignment, matching and identification. <i>Analytical chemistry</i>. Benjamini et al., 1995. Controlling the False Discovery Rate: A Practical and Powerful Approach to Multiple Testing. <i>Journal of the Royal Statistical Society</i>. Thévenot et al., 2023. phenomis: Postprocessing and univariate analysis of omics data. <i>Scientific data</i>. Gu et al., 2016. Complex heatmaps reveal patterns and correlations in multidim genomic data. <i>Bioinformatics</i>. 						
 1- INRAE: Laboratoire d'Immuno-Allergie Alimentaire (LIAA) 2- CEA: Laboratoire Innovations en spectrométrie de masse pour la santé (LIMS) 3- INSERM UMR1167, RID-AGE Université de Lille, Centre Hosp. Univ Lille, Institut Pasteur de Lille, INSERM, INRAE, Lille 4- Centre de Recherche en Epidémiologie et Statistiques (CRESS), Université de Paris, INSERM, INRAE, Paris 								