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Human leukocyte antigendependent colonization of *Lactobacillus* in the early-life gut

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To determine the importance of Lactobacillus in shaping the human gut microbiome, the microbial composition of stools from 1,602 children between the ages of 0.3 months and 37.2 months was analyzed in a general population cohort in the All Babies in Southeast Sweden study. Lactobacillus colonized only 32% of the total pediatric population at an average relative abundance of 0.29%. Lactobacillus was age-dependent, decreasing in prevalence and relative abundance over time. The main determining factor for Lactobacillus colonization was whether the individual was actively breastfeeding. Following cessation of breastfeeding, Lactobacillus prevalence rapidly declined. However, within the actively breastfeeding cohort, 45.6% of the population remained uncolonized by Lactobacillus. The presence versus absence of Lactobacillus was determined to be human leukocyte antigen (HLA) dependent. Individuals with HLA DR15-DQ6.2 were 3.4 times more likely to be colonized by Lactobacillus than those without the haplotype, and those with HLA DR5-DQ7 were more likely to have zero Lactobacillus despite actively breastfeeding. These results suggest that HLA genetics should be considered when designing Lactobacillus-based probiotics.

KEYWORDS

Lactobacillus, ABIS, microbiome, probiotic, type 1 diabetes, celiac disease, DR15-DQ6, DR5-DQ7

1 Introduction

Over a century ago, Metchnikoff espoused widely his observations that lactic acid fermenting microbes help prevent intestinal malady, and communities that regularly consumed fermented foods had extended longevity (Metchnikoff, 1908). Today, the study of lactic acid fermenting microbes is substantially driven by the probiotic industry —the global probiotics market size is expected to reach over \$100 billion by 2030—and, of the vast selection of probiotic supplements available, *Lactobacillus* and *Bifidobacterium* are the most common (Probiotics: What You Need To Know, 2019; Probiotics Market Size | Industry Report, 2021–2030, 2022). Despite the numerous probiotic research studies and

clinical trials published, results are frequently conflicting or inconclusive (Floch et al., 2015; Suez et al., 2019; Berryman et al., 2022). We previously showed in a meta-analysis of probiotic clinical trials for the treatment of autoimmune diseases that 48 of the 65 measurable results from 19 trials were not significant (Berryman et al., 2022). Of the 50 genera used in the examined trials, *Lactobacillus* was included in 42% of the supplementations and *Bifidobacterium* was included in 40%. Despite the probiotic industry's reliance on *Lactobacillus*, the importance and role that *Lactobacillus* plays in ameliorating dysbiosis remains unclear.

Inconclusive or non-significant results do not mean that the study of potential probiotic therapies is not valuable. Microbial dysbiosis has been associated with autoimmune diseases, neurologic diseases, and cardiovascular diseases (Angelakis et al., 2012; Carding et al., 2015; Russell et al., 2019; Milletich et al., 2022; Kundu et al., 2023; Park et al., 2023). Specifically, *Lactobacillus* enrichment and depletion have been seen in Crohn's disease, rheumatoid arthritis, obesity, type 2 diabetes, irritable bowel syndrome, type 1 diabetes, HIV, and multiple sclerosis (Heeney et al., 2018). *Lactobacillus* is a commensal bacterium inhabiting humans from birth and has been isolated from the oral cavity, gastrointestinal tract, skin, and vagina (Chu et al., 2017). While *Lactobacillus* can dominate the vaginal microbiome, it makes up \leq 1% of the adult human gut bacteria (Chu et al., 2017; Heeney et al., 2018).

Since previous studies of both adult and infant populations show that the average relative abundance of *Lactobacillus* in the gut of healthy individuals is quite low (Rossi et al., 2016; Almonacid et al., 2017; Chu et al., 2017; Beck et al., 2022), the objective of this study was to understand levels of *Lactobacillus* in early gut microbiome development and further investigate why a large percentage of the population is not colonized by the genus. We analyzed the microbial composition of stools at different ages from a large general population cohort of children to determine the genus' colonization patterns over time with respect to its microbial cohabitants. The prevalence of *Lactobacillus* in this population is remarkably low, even with a detection limit of 0.0077%. We hypothesized that *Lactobacillus* may be only a minor player in the human gut microbiome starting in infancy.

2 Results

2.1 Cohort description

This study analyzed the V3-V4 region of 16S rRNA in stool samples from the prospective, general-population cohort of the All Babies in Southeast Sweden (ABIS) study, which consisted of 17,055 children born between 1 October 1997 and 1 October 1999 in southeast Sweden (Ludvigsson et al., 2001). Biological specimens were longitudinally collected from birth to 13 years of age, creating a biobank of stool, urine, blood, and hair. Metadata was acquired in the form of regular questionnaires, first-year diaries, and human leukocyte antigen (HLA) genotyping. Parental report questionnaires and diaries included information about delivery mode, infection history, antibiotic use, duration of breastfeeding, introduction to and frequency of certain food consumption, living conditions, and additional environmental factors. Stool samples were analyzed from 1,602 children between the ages of 0.3 months and 37.2 months, with an average age of collection of 11.94 months (Supplementary Table 1). This analysis included one stool sample from each child, with a total of 1,602 samples analyzed in a between-subjects design. The average read count for the 1,602 samples was 72,113, the minimum read count was 13,183, and the maximum read count was 623,767. To normalize the data, the read counts for the 1,602 samples analyzed were rarefied to the minimum read count of 13,183. Total abundance as reads per gram was calculated by multiplying the relative abundance and copies of 16S rRNA per gram of stool found from qPCR (Jian et al., 2020).

2.2 *Lactobacillus* does not colonize the majority of pediatric guts

Upon analysis of the total population cohort, 1,602 children aged 0.3 months to 37.2 months at stool collection, it was determined that *Lactobacillus* did not colonize the majority of the pediatric population (Table 1). An average total abundance of 224,378 (\pm 4,007,910; min: 0; max: 156,941,755) *Lactobacillus* reads per gram of stool were detected, with an average relative abundance of 0.29% (\pm 2.3%; min: 0%; max: 74.5%) in the total population. Zero *Lactobacillus* reads were detected in 68.2% of the total population. For comparison, the average total abundance of *Bifidobacterium* was 139,896,70 (\pm 95,585,095; min: 0; max: 2.4e +09) reads per gram of stool, with an average relative abundance of 16.6% (\pm 18.8%; min: 0%; max: 98.9%) and 98.5% prevalence. *Lactobacillus* ranked 50th in prevalence out of the 205 genera found and populated 31.8% of the entire ABIS cohort studied here (Figure 1A).

Within the genus, 25 Lactobacillus amplicon sequence variants (ASVs), each with at least one nucleotide difference, were found within the total population. Every Lactobacillus ASV was detected in a small percentage of the cohort. Thirteen ASVs were in fewer than 1% of the population and only three were in more than 5% of the population, at 9.3%, 9.1%, and 8.2% prevalence, respectively (Figure 1B). The top three ASVs in more than 5% of the population were identified using the SILVA_v138 database as Lactobacillus NA 5972, Lactobacillus delbrueckii 5913, and Lactobacillus NA 5971. The 16S sequences from each ASV were individually searched with NCBI BLAST and verified or identified with 100% query coverage and \geq 99.7% identity. The top three ASVs were identified as Lacticaseibacillus paracasei, Lactobacillus delbrueckii, and Lacticaseibacillus rhamnosus (Supplementary Table 2). The frequency of possessing multiple Lactobacillus ASVs was low, with only 11.8% of the population colonized by more than one Lactobacillus ASV-321 children (20.0%) had one ASV, 96 (5.9%) had two ASVs, 44 (3.7%) had three ASVs, 25 (1.6%) had four ASVs, 15 (0.9%) had five ASVs, six (0.4%) had six ASVs, and three (0.2%) had seven ASVs. No individual was colonized by more than seven Lactobacillus ASVs.

TABLE 1 Prevalence of Lactobacillus absence vs. presence in total and subset cohorts.

		Lactobacillus	
	Cohort size	Absence	Presence
Total population	1,602 (100%)	1,092 (68.2%)	510 (31.8%)
Age: 0–3 months	16 (9.9%)	11 (68.7%)	5 (31.3%)
Age: 3–6 months	88 (5.5%)	40 (45.5%)	48 (54.5%)
Age: 6–12 months	543 (33.9%)	369 (68%)	174 (32.0%)
Age: 12-18 months	854 (53.3%)	610 (71.4%)	244 (28.6%)
Age: >18 months	27 (1.7%)	20 (74.1%)	7 (25.9%)
Actively breastfeeding	83 (5.1%)	37 (44.6%)	46 (55.4%)
Not breastfeeding	1,273 (79.4%)	898 (70.5%)	375 (29.5%)
DR15-DQ6.2 present	18 (1.1%)	4 (22.2%)	14 (77.8%)
DR5-DQ7 present	12 (0.7%)	10 (83.3%)	2 (16.7%)
DR8-DQ4 present	6 (0.4%)	5 (83.3%)	1 (16.7%)

Active breastfeeding: diet included breastmilk at the time of stool sample collection. Not breastfeeding: diet did not currently include breastmilk at the time of stool sample collection.

2.3 *Lactobacillus* is age-dependent and associated with breastfeeding

In the total population, *Lactobacillus* relative abundance and total abundance were negatively correlated with increasing age calculated with Spearman's rank correlation (RA: rho = -0.096, p = 1.6e-4; TA: rho = -0.10, p = 8.9e-5). In addition, whether a child was actively breastfeeding when their stool sample was provided was the only variable significantly associated with both *Lactobacillus* relative abundance and/or total abundance when adjusted for false discovery

rates using the Benjamini–Hochberg method (Kruskal–Wallis: p_{adj} = 2.4e-17 and p_{adj} = 4.2e-18, respectively) (Figure 2). An individual was categorized as actively breastfeeding if their diet included breastmilk, either exclusively or supplementally, at the time of stool sample collection. An individual was categorized as not breastfeeding if their diet did not include breastmilk at the time of stool sample collection.

Further investigation into the defining features of those who were colonized by *Lactobacillus*, referred to here as the *Lactobacillus* presence cohort (LPC), was performed. The average total abundance of *Lactobacillus* in the LPC was 705,820 (\pm 7,089,251;





min: 4; max: 156,941,755) reads per gram of stool, with 0.92% (\pm 3.9%; min: 0.0075%; max: 74.5%) average relative abundance. As with the total population in the LPC, *Lactobacillus* relative abundance decreased significantly with age (Figure 3). In the 0 months to 6 months LPC, the mean relative abundance of *Lactobacillus* was 2.3% (\pm 10.7%). In the 6 months to 12 months LPC, the mean relative abundance of *Lactobacillus* was 0.79% (\pm 2.0%). In the 12 months to 18 months LPC, the mean relative abundance of *Lactobacillus* was 0.51% (\pm 1.4%). In the over 18 months LPC, the mean relative

abundance of *Lactobacillus* was 0.02% (\pm 0.01%). The relative abundance of *Bifidobacterium* was shown over time for comparison: it significantly declines after 6 months but remains steady. *Lactobacillus* continued to significantly decrease as age increased. Through a chi-squared test of independence, *Lactobacillus* presence versus absence was most dependent on active breastfeeding ($p_{adj} = 0.028$) (Supplementary Table 3).

The prevalence of *Lactobacillus* within the total population increased during the period from birth to 6 months, a time



period when the majority of the population was actively breastfeeding (Figure 4). Despite the majority of the population continuing to receive breastmilk up to 9 months, the prevalence of Lactobacillus colonization in the total population began to decline after 6 months and dropped starkly after 7 months. The 5 months to 6 months age group had the highest prevalence of Lactobacillus colonization at 61.5%. After 6 months, the prevalence of Lactobacillus colonization dropped from 53.8% at age 6-7 months to 33.3% at age 7-8 months, and 29.1% by 10-11 months. By 17-18 months, Lactobacillus colonization was at 11.1% prevalence.

2.4 Human leukocyte antigen associated with Lactobacillus presence in actively breastfeeding cohort

While actively breastfeeding was a major determinant of Lactobacillus presence, relative abundance, and total abundance, only 55.4% of the total actively breastfeeding portion of the cohort (ABC) was colonized by Lactobacillus (Table 1). Since 45.6% of the ABC remain uncolonized by Lactobacillus, statistical analysis was performed to investigate which variables impact whether an actively breastfeeding child is colonized or not. It was determined that microbial factors did not influence Lactobacillus colonization in the ABC. No other genera relative abundances or total abundances were significantly associated with Lactobacillus presence or absence, when Wilcoxon rank-sum tests were performed and p-values were adjusted for false discovery rate (data not shown). There were also no strong correlations between Lactobacillus relative abundance and other genera with Spearman's rank correlations. Subsequently, it was determined that additional environmental factors did not affect the colonization of Lactobacillus in the ABC. Chi-squared tests and a Kruskal-Wallis one-way analysis of variance of the environmental factors in parental report questionnaires and diaries showed no significant association with Lactobacillus presence versus absence or relative abundance (data not shown).

Human leukocyte antigen genetics was the only factor tested that showed a significant relationship with whether an actively breastfeeding infant was colonized by Lactobacillus. Lactobacillus presence was determined to be dependent on HLA haplotypes DR5-DQ7 (p = 0.0045), DR15-DQ6.2 (p = 0.048), and DR8-DQ4 (p = 0.048) 0.078) through a chi-squared test of independence. An odds ratio analysis indicated that individuals with HLA DR15-DQ6.2 were 3.4 times more likely to be colonized by Lactobacillus than those without that haplotype. Individuals with HLA DR5-DQ7 were more likely to not be colonized by Lactobacillus despite actively breastfeeding (Figure 5).

3 Discussion

Our analysis of 1,602 pediatric microbiomes over time reaffirms that Lactobacillus is not a major component of the gut after 6 months of age. At an average age of 11.94 months, nearly 70% of the total population was not colonized by Lactobacillus at all. Of those who were colonized by Lactobacillus, the relative abundance decreased steadily over time. For most of the population, Lactobacillus did not remain in the gut following the cessation of breastfeeding. Breastfeeding was the major determinant of whether an individual was colonized by Lactobacillus or not and the relative and total abundances an individual had. Controlling for active breastfeeding revealed HLA associations that had previously been masked. Individuals with HLA DR15-DQ6.2 were 3.4 times more likely to be colonized by Lactobacillus than those without the haplotype, and those with HLA DR5-DQ7 were more likely to have zero Lactobacillus despite actively breastfeeding.



whose diet included breastmilk at the time of stool sample collection. There were no stool samples collected at age 8–9 months



Our results confirm that the low amounts of *Lactobacillus* colonizing the human gut start in infancy. Previous studies of both adult and infant populations show that the average relative abundance of *Lactobacillus* in healthy individuals without supplementation is ~0.3%, which matches our finding of the 0.29% average relative abundance in the pediatric population studied here (Rossi et al., 2016; Almonacid et al., 2017; Chu et al., 2017; Beck et al., 2022). While a general population study of human adults found 43 *Lactobacillus* strains consistently residing in the gut (Rossi et al., 2016), this study observed only 25 *Lactobacillus* ASVs within the pediatric population, which may be a result of exposure due to age.

Lactobacillus and Bifidobacterium are predominant composites of the human breastmilk microbiome (Łubiech and Twarużek, 2020). It is unsurprising that both Lactobacillus and Bifidobacterium levels are higher up to 6 months of age, which is when the cohort was primarily breastfeeding. However, at our levels of detection, Lactobacillus did not appear to remain in most guts following the cessation of breastfeeding. Lactobacillus abundance and prevalence increased at 3 months and declined after 6 months as the population began to cut back on breastfeeding. The persistent decline in Lactobacillus relative abundance over time was not observed in the other breastfeedingassociated genus, Bifidobacterium. The lack of persistent Lactobacillus after supplementation was consistent with a recent infant probiotic study that showed that following treatment with Bifidobacteriumand Lactobacillus-based probiotics, some Bifidobacterium persisted but no Lactobacillus persisted in the infant gut (Beck et al., 2022). Bifidobacterium was noticeably more abundant than Lactobacillus in the cohort studied here, which is supported by previous studies (Vatanen et al., 2018).

When controlling for active breastfeeding, an association with interesting implications was revealed. Individuals with HLA DR15-DQ6.2 were more likely to be colonized by *Lactobacillus* and those with HLA DR5-DQ7 were more likely to have zero *Lactobacillus* despite actively breastfeeding. The influence that active supplementation of *Lactobacillus* through breastfeeding had on the gut microbiome composition likely masked the influences of genetics during initial investigations. Still, almost half of the actively breastfeeding population in this study was not colonized by *Lactobacillus* at all. The determining factor for this lack of colonization was possessing HLA DR5-DQ7. To our knowledge, the negative association between HLA DR5-DQ7 and *Lactobacillus* has not been previously reported. However, DR5-DQ7 is the third most abundant haplotype in celiac disease patients, often seen in a heterozygote pair with DR7-DQ2 (Tinto et al., 2015; Pisapia et al., 2020). Together with evidence that *Lactobacillus* is seen in lower abundance in celiac disease than in healthy control patients (Lorenzo Pisarello et al., 2015), this suggests that HLA is playing an important role in shaping the microbiome. HLA DR15-DQ6.2 is a known protective allele against type 1 diabetes (Thomas et al., 2021), and its link here with *Lactobacillus* further emphasizes the need to carefully consider autoimmune-associated genetics when studying probiotic bacteria.

Lactobacillus rhamnosus was among the highest Lactobacillus ASV detected in this study. Interestingly, L. rhamnosus is higher in healthy controls and may be protective against islet autoimmunity (Vatanen et al., 2018). However, Lactobacillus was included in 42% of the supplementations in 19 probiotic clinical trials for the treatment of autoimmune disease over the past 10 years, and the majority of the results were not significant. Specifically, L. rhamnosus was included in all three of the T1D probiotic studies and none of them showed prevention or amelioration of the disease (Berryman et al., 2022). None of these studies controlled for genetics, however. The potential for success may be found if future studies take HLA into consideration. While Lactobacillus is arguably poor at colonizing the gut, a previous study concluded that heat-inactivated, noncolonizing Lactobacillus strains were sufficient in promoting the regulation of intestinal epithelial barrier function (Singh et al., 2021), suggesting that further analysis needs to be conducted to determine if the low colonization level of Lactobacillus is an inhibiting factor when acting as a probiotic.

Lactobacillus is a minor player in the human gut microbiome starting in infancy, even during breastfeeding. It does not readily colonize the gut without active and persistent intake. As a potential component of probiotic supplementation, it is important to consider the role HLA may play in promoting or preventing colonization of *Lactobacillus*. If autoimmune risk-associated HLAs are inhibiting the colonization of potentially beneficial bacteria, it could indicate that the individuals who could benefit most from the amelioration of dysbiosis are not being treated efficiently by the probiotics available. Future research is needed to determine the best probiotic formula for high-risk individuals.

4 Materials and methods

4.1 Sample collection and institutional review board approval

The All Babies in Southeast Sweden (ABIS) study included a prospective general population cohort of 17,055 children born in southeast Sweden between 1 October 1997 and 1 October 1999. Parental consent followed oral, written, and video information about the study. Following parental consent, participating parents completed questionnaires and diaries from birth through the first year of life (Ludvigsson et al., 2001). Collected information included but was not limited to infection history, antibiotic use, duration of breastfeeding, introduction to or frequency of certain food consumption, living conditions, and additional environmental factors. Approval was obtained by the Research Ethics Committees of the Faculty of Health Science at Linköping University, Sweden, Ref. 1997/96,287 and 2003/03-092, and the Medical Faculty of Lund University, Sweden (Dnr 99227, Dnr 99321) as described previously (Russell et al., 2019). The microbial analysis performed at the University of Florida was approved by the University of Florida's Institutional Review Board as an exempt study IRB201800903.

Pediatric stool samples were collected from 1,756 individual children aged 0.3 months to 37.2 months—the average age of collection was 11.94 months—and processed as previously reported (Russell et al., 2019). Following filtering, this analysis included one stool sample from each child, with a total of 1,602 samples analyzed in a between-subject effect design.

All available HLA genotypes determined from blood spots were included in the analysis. HLA DR-DQ haplotypes of interest here were defined as follows: DR5-DQ7 (*DRB1*11-DQA1*05-DQB1*03*); DR8-DQ4 (*DRB1*08-DQA1*03:03-DQB1*04*); and DR15-DQ6.2 (*DRB1*15-DQA1*01:02-B1*06:02*). Typing was performed with sequence-specific lanthanide-labeled oligonucleotide hybridization (Ilonen et al., 2016).

4.2 Processing amplicon sequence variants

DNA extraction, 16S rRNA barcoded PCR, and V3-V4 16S rRNA Illumina sequencing were performed on stool samples collected as previously described (Russell et al., 2019). Universal 16S rRNA primers were used to perform bacterial quantification through quantitative polymerase chain reaction (qPCR) as previously described (Russell et al., 2021). Paired-end reads were merged, primers were removed, and ASV processing was performed as previously described (Milletich et al., 2022). Taxonomy was assigned via the assignTaxonomy() function using the SILVA_v138 database within the R package DADA2 (version 1.26) (Pruesse et al., 2007; Quast et al., 2013; Callahan et al., 2016). Species assignments for Lactobacillus ASVs that were below default minimum boostrapping confidence were determined via NCBI BLAST with a 100% query coverage and \geq 99.7% identity (Supplementary Table 2) (Altschul et al., 1990; Boratyn et al., 2012).

Filtering the original 1,756 children to exclude samples with fewer than 1,000 total reads resulted in the analysis of stool samples from 1,602 children. Filtering to remove ASVs with fewer than five reads using the filter_taxa function from phyloseq (version 1.42.0) resulted in 2,102 ASVs (McMurdie and Holmes, 2013). The average read count for the 1,602 samples analyzed was 72,113, the minimum read count was 13,183, and the maximum read count was 623,767. To normalize the data, the read counts for the 1,602 samples analyzed were rarefied to the minimum read count of 13,183 using rarefy_even_depth() from the phyloseq package. The relative abundance of samples was determined with transform_sample_counts, and the reads/g was calculated by multiplying the relative abundance and copies of 16S rRNA per gram of stool found from qPCR (Jian et al., 2020). ASVs were conglomerated into genera using the tax_glom() function from phyloseq.

4.3 Statistical analysis

All packages were run on RStudio (version 2023.03.0 + 386) (Posit team, 2023). Variables impacting binomial beta diversity were tested for using the permutational multivariate analysis of variance (PERMANOVA) test through the adonis() function in the package vegan (version 2.4-6) (Oksanen et al., 2020). Alpha diversity was calculated with the R function plot_richness() in the phyloseq package. Confounding factors of *Lactobacillus* abundance and relative abundance were determined via a non-parametric Kruskal-Wallis rank sum test using the R function kruskal.test(). Genera impacted by the presence or absence of *Lactobacillus* were determined via the non-parametric Wilcoxon rank sum test using the R function wilcox.test(), and *p*-values were corrected for false discovery rates (FDRs) using the Benjamini–Hochberg method using the R function p.adjust().

The R package PIME (version 0.1.0) determined core microbiomes of each cohort (Roesch et al., 2020). The plot_ordination() function constructed the associated principal coordinate analysis (PCoA) plot.

Relative abundances were calculated using the transform _sample_counts function. To calculate total abundance, the relative abundance values were multiplied by the total number of copies of 16s rRNA per gram of stool, as determined through qPCR. Prevalence was defined as the percentage of individuals in the cohort with non-zero abundance of the respective ASV or taxa. Prevalence was calculated using the getPrevalence() function from R package mia (version 1.4.0) (Ernst et al., 2023).

The odds ratio was calculated with the logistic regression model R function glm() and odds.ratio() function from the questionr package (version 0.7.8). The Spearman's correlation was calculated with the R function cor(). The correlation plot was created using the R package corrplot (version 0.92) (Wei and Simko, 2023). Odds ratios were calculated using the oddsratio() function from the R package epitools (version 10.1) and MedCalc Software Ltd (Aragon et al., 2020; Schoonjans, n.d). The odds ratio plot was created using the or_plot() function from the finalfit package (version 1.0.6) (Harrison et al., 2023). The heatmaps were designed using the R package pheatmap (version 1.0.12) (Kolde, 2019). Boxplots were designed using the R package

ggplot2 (version 3.4.0) (Wickham, 2016). The function stat_compare_means() was used to determine the Kruskal–Wallis and Wilcoxon p-values. The function ggarrange() was used to arrange each graph into a figure.

Data availability statement

The original contributions presented in the study are included in the Supplementary Material Data 1 and Data 2, further inquiries can be directed to the corresponding author.

Ethics statement

The studies involving humans were approved by Research Ethics Committees of the Faculty of Health Science at Linköping University, Sweden, Ref. 1997/96287 and 2003/03-092 and the Medical Faculty of Lund University, Sweden (Dnr 99227, Dnr 99321). The studies were conducted in accordance with the local legislation and institutional requirements. Written informed consent for participation in this study was provided by the participants' legal guardians/next of kin.

Author contributions

MB developed the concept, performed the primary data analysis, and wrote the paper. ET assisted with concept development and interpretation of data. JL founded and coordinated ABIS, designed the study, and carried out sample collection, storage, and transport. All authors contributed to the article and approved the submitted version.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/frmbi.2023.1192773/ full#supplementary-material

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