

4 Human milk oligosaccharide DSLNT and gut microbiome in preterm infants predicts necrotising
5 enterocolitis
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33

34 **List of abbreviations**

35 NEC - Necrotising enterocolitis

- 36 MOM - Mothers own milk, MOM
- 37 HMO - Human milk oligosaccharide
- 38 2'FL - 2'-fucosyllactose
- 39 3FL - 3-fucosyllactose
- 40 LNnT - Lacto-N-neotetraose
- 41 3'SL - 3'-sialyllactose
- 42 DFlac - Difucosyllactose
- 43 6'SL - 6'-sialyllactose
- 44 LNT - Lacto-N-tetraose
- 45 LNFP - Lacto-N-fucopentaose
- 46 DFLNT - Difucosyl-LNT
- 47 LNH - Lacto-N-hexaose
- 48 DSLNT - Disialyllacto-N-tetraose
- 49 FLNH - Fucosyl-lacto-N-hexaose
- 50 DFLNH - Difucosyl-lacto-N-hexaose
- 51 FDSLNH - Fucosyl-disialyl-lacto-N-hexaose
- 52 DSLNH - Disialyl-lacto-N-hexaose
- 53 PGCT - Preterm gut community types
- 54 DOL - Day of life
- 55 IQR - Interquartile range
- 56 ROC - Receiver operating characteristic
- 57 SVM - Support Vector Machine
- 58 MCCV - Monte-Carlo cross validation
- 59 PERMANOVA - Permutational multivariate analysis of variance
- 60 MDA - Mean decrease accuracy
- 61 DMM - Dirichlet multinomial modelling
- 62 PMA - Postmenstrual age
- 63 NICU - Neonatal intensive care unit

64 **Abstract**

65 Objective: Necrotising enterocolitis (NEC) is a devastating intestinal disease primarily affecting
66 preterm infants. The underlying mechanisms are poorly understood: mothers own breast milk
67 (MOM) is protective, possibly relating to human milk oligosaccharide (HMO) and infant gut
68 microbiome interplay. We investigated the interaction between HMO profiles and infant gut
69 microbiome development and its association with NEC.

70 Design: We performed HMO profiling of MOM in a large cohort of infants with NEC (n=33) with
71 matched controls (n=37). In a subset of 48 infants (14 NEC) we also performed longitudinal
72 metagenomic sequencing of infant stool (n=644).

73 Results: Concentration of a single HMO, disialyllacto-N-tetraose (DSLNT), was significantly lower
74 in MOM received by NEC infants compared to controls. A MOM threshold level of 241 nmol/mL
75 had a sensitivity and specificity of 0.9 for NEC. Metagenomic sequencing before NEC onset
76 showed significantly lower relative abundance of *Bifidobacterium longum* and higher relative
77 abundance of *Enterobacter cloacae* in infants with NEC. Longitudinal development of the
78 microbiome was also impacted by low MOM DSLNT associated with reduced transition into
79 preterm gut community types dominated by *Bifidobacterium* spp. and typically observed in older
80 infants. Random forest analysis combining HMO and metagenome data before disease accurately
81 classified 87.5% of infants as healthy or NEC.

82 Conclusion: These results demonstrate the importance of HMOs and gut microbiome in preterm
83 infant health and disease. The findings offer potential targets for biomarker development, disease
84 risk stratification, and novel avenues for supplements that may prevent life-threatening disease.

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90 **1. What is already known about this subject?**

- 91 • Necrotising enterocolitis (NEC) is one of the leading causes of death in preterm infants
- 92 • Maternal human milk oligosaccharides (HMOs) including disialyllacto-N-tetraose (DSLNT)
- 93 have been associated with protection from NEC development
- 94 • Differences in infant gut microbiome development have been linked to NEC and non-NEC
- 95 infants, but the causative and protective organisms have not been determined

96

97 **2. What are the new findings?**

- 98 • We found for the first time that combined analysis of maternal HMOs and infant gut
- 99 microbiome can predict NEC
- 100 • A specific DSLNT threshold level of 241 nmol/mL had a sensitivity and specificity of 0.9
- 101 for NEC and infants receiving milk below this threshold showed abnormal microbiome
- 102 development
- 103 • Infants who developed NEC had significantly lower relative abundance of *Bifidobacterium*
- 104 *longum* and significantly higher relative abundance of *Enterobacter cloacae* before disease
- 105 diagnosis

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107 **3. How might it impact on clinical practice in the foreseeable future?**

- 108 • Our findings demonstrate the importance of maternal HMOs and infant gut microbiome in
- 109 preterm infants, providing targets for biomarker development, disease risk stratification, and
- 110 novel avenues for supplementing the infant feed.

111

112 **Introduction**

113 Necrotising enterocolitis (NEC) is an inflammation-mediated bowel condition that is a leading
114 cause of death and serious morbidity in preterm infants born before 32 weeks gestation (1). The
115 mechanisms underlying NEC development are poorly understood and the lack of specificity of
116 symptoms and tests make diagnostic certainty difficult. Infants with NEC have enteral feeds
117 stopped and are treated with broad-spectrum antibiotics, and may need surgery (2).

118 Receipt of mother's own breast milk (MOM) is the most protective factor against the development
119 of NEC in preterm infants (3, 4). However, infants receiving MOM still develop NEC, suggesting
120 the variable composition of nutrients and other components of breast milk may be important.

121 Human milk oligosaccharides (HMOs) are structurally diverse, complex unconjugated sugars that
122 are not usually present in artificial formula milk (5). HMOs are indigestible to the infant, reaching
123 the lower gastrointestinal tract intact where they act as growth substrates (i.e. prebiotics) for specific
124 bacteria, notably *Bifidobacterium* spp. thought key to infant health (6-8). HMOs may also protect
125 from enteric organism blood stream infections due to anti-microbial activity (9), stimulate the
126 immune system (10), enhance gut barrier function (11), and act as decoy receptors for pathogens
127 (12). While >150 HMOs have been described, the 19 most abundant represent >95% of the total
128 HMO content (13). HMO profiles are specific to individual mothers and remain relatively stable
129 during lactation (14). Presence of an active FUT2 gene, which is involved in the synthesis of α 1-2-
130 fucosylated oligosaccharides, is the main determinant of the HMO profile, termed maternal secretor
131 status (15).

132 Recent work has begun to elucidate the potential contribution of HMOs to preterm infant health. In
133 a neonatal rat model, disialyllacto-N-tetraose (DSLNT), a non-fucosylated, but double-sialylated
134 HMO, significantly reduced NEC development and improved NEC-associated mortality rate (16).

135 An association of lower DSLNT concentration in MOM and subsequent higher risk of NEC onset in
136 the infant has since been observed in preterm human studies (17-19). To date, these studies have

137 included very small numbers of infants with NEC (between 4 and 8), with a broad range of NEC
138 phenotypes. Thus, validation in a larger cohort is urgently needed.

139 Altered gut microbiome development has been associated with NEC in preterm infants. While no
140 specific causative microorganism has consistently been identified, studies have reported a higher
141 relative abundance of Enterobacteriaceae, coupled to lower relative abundance of *Bifidobacterium*
142 (20-23). Instability of the gut microbiome in NEC infants has also been reported in longitudinal
143 studies, with more frequent transitions between different preterm gut community types (PGCT) in
144 NEC (20). These findings were replicated at the site of disease in a study using formalin fixed
145 paraffin embedded tissue from NEC infants matched to non-NEC controls (24). Previous
146 microbiome studies have largely relied on 16S rRNA gene sequencing of the V4 region, which has
147 limited resolution, especially for emerging key organisms of interest for preterm health (i.e.,
148 *Klebsiella* and *Enterobacter* would be classified together as Enterobacteriaceae). Metagenomics
149 may overcome this and recent metagenomic data showed infants who developed NEC had higher
150 relative abundance of *Klebsiella* and higher replication rates in all bacteria before disease onset
151 (25).

152 In this current study, we performed a combined analysis of maternal HMO profiles and longitudinal
153 development of the infant stool/gut microbiome in a large cohort of preterm infants with NEC and
154 healthy controls matched for gestation, birthweight and day of life. We then validated our results in
155 an independent cohort using previously published HMO data (17). We hypothesised that differences
156 in maternal HMO profiles and microbiome development may explain why some infants receiving
157 MOM still develop NEC.

158

159 **Methods**

160 **Cohort**

161 This study included 77 preterm infants (born at <32 weeks gestation), who were born in or
162 transferred to a single large tertiary level neonatal intensive care unit (NICU) in Newcastle upon
163 Tyne, UK recruited to the SERVIS study (REC10/H0908/39) with written parental consent
164 covering data and sample collection. 33 infants were diagnosed with definite NEC and 37 non-
165 diseased controls were selected by identifying a healthy infant matched by gestation, birthweight,
166 and having a MOM sample available at a corresponding day of life (DOL) (**table 1**). Detailed
167 information on feeding and antibiotic use are included in **online supplementary table 1 and online**
168 **supplementary table 2**. Diagnoses were made using an extensive combination of clinical, x-ray
169 and histological findings and blindly agreed by two neonatal clinicians (JEB and NDE). Standard
170 clinical protocols recommended the routine use of supplemental probiotics when more than
171 30mL/kg/day of MOM was tolerated for at least 1-2 days: all 33 NEC infants received MOM and
172 31 received probiotics. All 37 controls received probiotics. The probiotics administered were either
173 LaBiNIC (*Lactobacillus acidophilus*, *Bifidobacterium infantis* and *Bifidobacterium bifidum*) or
174 Infloran (*L. acidophilus* and *B. bifidum*).

175 The median DOL of NEC diagnosis was 19 (interquartile range; IQR 14-35; **table 1**). A single
176 MOM sample was analysed for each infant, as close to the onset of disease as possible, with control
177 samples matched by day of life (DOL) (**online supplementary figure 1**). The median DOL of
178 MOM from NEC cases was 18 (IQR 13-34) and from controls was also 18 (IQR 12-31). The DOL
179 of the milk sample is the DOL received by the infant, and is not necessarily the same day as the
180 mother expressed the milk due to standard practice that often involves milk storage. Metagenomic
181 sequencing of stool samples (n = 644) was performed longitudinally on a subset of 48 infants
182 (including 14 NEC; **online supplementary figure 1**). These infants were comparable to the full
183 cohort (**online supplementary table 3**).

184

185 Full details of the HMO, metagenome, and statistical analysis are described in **online**
186 **supplementary methods.**

187

188 *Human milk oligosaccharides analysis*

189 The absolute quantification for the 19 most abundant HMOs was determined by high-performance
190 liquid chromatography (HPLC) following derivatization as per the protocol described by Bode *et al.*
191 (26). Maternal secretor (presence of an active FUT2 gene) status was determined by presence or
192 near-absence of 2'FL in the breast milk analysed.

193

194 *Metagenomes*

195 DNA was extracted from ~0.1g of stool using the DNeasy PowerSoil Kit (QIAGEN) following the
196 manufacturer's protocol and sequencing was performed on the HiSeq X Ten (Illumina) with a read
197 length of 150bp paired end reads. Processed fastq files were mapped against the MetaPhlan2 marker
198 gene database (mpa_v20_m200) (27).

199

200 *Statistical analysis*

201 Statistical analysis of HMO profiles was performed using MetaboAnalyst 3.0 (28). For ordinations,
202 HMO data was normalised by logarithmic transformation and 2000 random permutations were used
203 to test the significance. Multivariate ROC curves were generated using linear Support Vector
204 Machine (SVM) classification method coupled with Monte-Carlo cross validation (MCCV).

205 Correlation between clinical variables and individual HMOs was tested by performing a
206 multivariate adjusted linear model in R (version 3.6.3). HMO concentrations were normalised by
207 log-transformation prior to analysis and P values were adjusted applying the Benjamini & Hochberg
208 correction (29).

209 The cross-sectional cohort of stool samples collected from NEC infants before diagnosis and
210 matched controls was analysed using MicrobiomeAnalyst (30, 31). Permutational multivariate

211 analysis of variance (PERMANOVA) was used to determine significance of Bray-Curtis principal
212 coordinate analysis. MetagenomeSeq was used to assess differential abundance at the phyla and
213 species level.

214 DMM clusters samples on the basis of microbial community structure (32) and was used to
215 determine the preterm gut community types (PGCTs) from all samples, as performed previously
216 (33, 34). Five PGCT was found to be optimal, and these were ordered youngest (PGCT-1) to oldest
217 (PGCT-5) based on the average DOL of samples within each PGCT. Analysis was performed at
218 specific time windows, including only a single sample per infant in each time point.

219 The association of various clinical variables on the HMO and metagenome profiles was tested by
220 applying the function “adonis” of “vegan” (version 2.5-6) package (35) in R, based on Bray-Curtis
221 dissimilarity and 10000 permutations. Each test was performed stepwise and P values were adjusted
222 using Benjamini & Hochberg (29).

223 Random Forest was used for comparing the performance of classification models built using
224 matched cross-sectional datasets.

225

226 **Results**

227 *Association of maternal HMOs and development of NEC in the infant*

228 MOM samples clustered according to maternal secretor status and secretor mothers had a higher
229 total HMO concentration, a higher HMO Shannon diversity, and a significantly higher
230 concentration of overall HMO-bound fucose (**online supplementary figure 2**). Thus, where
231 relevant, we have stratified and adjusted for maternal secretor status in subsequent analyses.

232 HMO profiles showed significant separation of NEC and control infants (**figure 1a**; 2000
233 permutations, $P < 0.001$) and this was consistent when secretor and non-secretor samples were
234 analysed separately (both $P < 0.001$; **online supplementary figure 3a and 3b**). Individually, of the
235 19 HMOs quantified in this study, only DSLNT was significantly different between NEC and
236 controls, with a lower concentration in NEC infants (adj. $P < 0.001$; **figure 1b and 1c**). No

237 significant associations were found in the Shannon diversity of HMOs between NEC and matched
238 controls for the full cohort, or when stratified by maternal secretor status (all $P > 0.05$; **online**
239 **supplementary figure 4**).

240 Given that lower DSLNT was associated with NEC independent of secretor status, the utility of this
241 HMO as a biomarker for NEC development was explored. Univariate ROC curve analysis
242 determined that 241 nmol/mL (or 310.93 $\mu\text{g/mL}$) was the optimal DSLNT concentration in MOM
243 for distinguishing NEC and control infants (**figure 1c** and **1d**). At this threshold, the area under the
244 curve (AUC) was 0.946 with a sensitivity of 0.9 and a specificity of 0.9, correctly identifying 91%
245 of NEC infants (below threshold) and 86% of control healthy infants (above threshold).

246 To test if integration of additional HMOs could improve the classification performance,
247 multivariate ROC curves built on increasing number of HMOs were performed (**online**
248 **supplementary figure 5a**). Inclusion of 2 HMOs (the minimum in multivariate analysis) resulted in
249 the optimal performance, with DSLNT being selected as a discriminatory feature in 100% of
250 permutations (**online supplementary figure 5b**). 3FL and LNnT were the 2nd and 3rd most selected
251 features, with a selection frequency of around 30%, being more abundant in cases of NEC.
252 However, the integration of any additional HMOs to DSLNT in the multivariate model resulted in
253 minimal improvement in performance compared to the univariate model using DSLNT only (AUC
254 of 0.949 and 0.946, respectively).

255 To validate the 241 nmol/mL threshold defined in the current study in an independent cohort, we
256 analysed data from Autran *et al.* (2018) which contained 8 NEC and 40 matched control infants
257 (17). Since this study included temporal sampling before disease, we selected the nearest milk
258 sample to NEC onset for each infant and matched the control samples by sample DOL and included
259 only DSLNT concentration. Using a DSLNT threshold of 241 nmol/mL, the MOM sample for
260 100% (8/8) NEC infants fell under the threshold, while 60% (24/40) control samples had a DSLNT
261 concentration above 241 nmol/mL (**online supplementary figure 5c**).

262

263 *Analysis of HMO profiles stratified by NEC type*

264 We compared medically managed NEC (NEC-M), where infants did not undergo surgery or die
265 from NEC (i.e. had less severe disease), with NEC infants that underwent surgery (NEC-S). NEC-
266 M and NEC-S clustered together and were distinct from matched controls (**figure 2a**; 2000
267 permutations, $P < 0.001$). Two HMOs were found to be significantly different, with DSLNT lower
268 in MOM in both NEC-M (adj. $P < 0.001$) and NEC-S (adj. $P < 0.001$) compared to controls (**figure**
269 **2b**). In addition, LNnT in MOM was significantly lower in NEC-S in comparison to both NEC-M
270 (adj. $P = 0.0016$) and matched controls (adj. $P = 0.0423$) (**figure 2c**).

271 We subsequently investigated the potential association between DSLNT and LNnT concentrations
272 and clinical variables by applying an adjusted linear model. DSLNT was negatively correlated to
273 both disease types, with coefficients equal to -0.60 for NEC-M (adj. $P < 0.001$) and -0.67 for NEC-
274 S (adj. $P < 0.001$) (**figure 2d**). However, LNnT was not associated with disease type following
275 adjusted linear modelling (both adj. $P > 0.05$). DSLNT and LNnT were both significantly higher in
276 secretor mothers (adj. $P = 0.008$, adj. $P < 0.001$, respectively). DSLNT in MOM also positively
277 correlated to gestational age (adj. $P = 0.008$) and negatively to birthweight (adj. $P = 0.008$). Neither
278 HMO correlated to sex, delivery mode, post-menstrual age, or DOL of the MOM sample (**figure 2d**
279 and **online supplementary figure 6**).

280

281 *Association of infant gut microbiome and development of NEC*

282 We included stool microbiome data on a subset of infants with HMO data, where metagenomic
283 sequencing data was available through an on-going independent study (the results of which are not
284 yet published). This included 644 stool samples from 34 controls, and 14 NEC infants (**online**
285 **supplementary table 3**). To overcome challenges of repeated measures and to compare results with
286 existing published work, we first analysed one stool sample per infant closest to NEC onset (median
287 of 3 days before NEC) and a corresponding control sample matched by DOL (**online**
288 **supplementary figure 1**). This cross-sectional analysis showed NEC infants had significantly

289 lower richness ($P = 0.027$) but comparable Shannon diversity ($P = 0.443$; **figure 3a**). Bray-Curtis
290 PCoA showed no significant difference between the bacterial profiles of NEC and controls
291 (PERMANOVA $P = 0.182$; **figure 3b**). Analysis at the phylum level showed significantly lower
292 relative abundance of Actinobacteria (adj. $P = 0.034$) and higher relative abundance of
293 Proteobacteria (adj. $P = 0.034$) in NEC infants (**figure 3c**). Correspondingly, at the species level,
294 NEC infants had lower relative abundance of *Bifidobacterium longum* (adj. $P = 0.012$) and higher
295 relative abundance of *Enterobacter cloacae* (adj. $P = 0.012$), compared to controls (**figure 3d**).

296

297 *Integrated analysis of HMO and bacterial profiles*

298 DMM clustering was used to determine preterm gut community types (PGCT) using species level
299 data and five PCGTs was deemed optimal (**figure 4a**). PGCT-1 was characterised by high relative
300 abundance of *Staphylococcus* spp. and *Enterococcus faecalis*, PGCT-2 had high *Escherichia* spp.,
301 PGCT-3 had high *Klebsiella* spp., and PGCT-4 and PGCT-5 had high *Bifidobacterium* spp. with *B.*
302 *breve* notably high in PGCT-5. Using the PGCT clusters, we analysed the temporal transition of an
303 infant's gut microbiome over the first 70 days of life by defining distinct time points and including
304 only one sample per infant at each time point. Based on the distribution of samples across all time
305 points and all clusters, the temporal transition of the microbiome over the first 70 days of life was
306 significantly different in infants in receipt of MOM below the DSLNT threshold of 241 nmol/ml
307 compared to infants above the DSLNT threshold (χ^2 test $P < 0.001$; **figure 4b**).

308 The PGCTs were named according to the average age of samples within that cluster, where PGCT-1
309 contained on average the earliest samples and PGCT-5 on average the latest samples. We compared
310 the number of samples from all time points in only PGCT-1 and PGCT-5 to investigate associations
311 between the MOM DSLNT threshold and gut microbiome development from the typically younger
312 to the typically older PGCTs. Infants receiving MOM with DSLNT level below 241 nmol/mL had
313 significantly more samples remaining within PGCT-1 throughout all time points (78% in PCGT-1
314 vs. 22% in PGCT-5, χ^2 test $P < 0.001$), whereas infants receiving MOM with DSLNT above this

315 threshold transitioned from PGCT-1 to PGCT-5 as demonstrated by a similar number of samples in
316 each PGCT across all time points (48% in PCGT-1 vs. 52% in PGCT-5, χ^2 test P = 0.717). In
317 addition to comparing samples from all times points, we next compared samples from the final time
318 point only (i.e., DOL 50-60). After correcting for uneven frequency of sampling between groups, at
319 the final time point infants receiving MOM above the DSLNT threshold were twice as likely to be
320 in PGCT-5 (3/11 samples below vs. 12/22 samples above DSLNT threshold; odds ratio 3.20, 95%
321 CI 0.6657 to 15.3819), which was characterised by high relative abundance of *Bifidobacterium*
322 (**figure 4a**).

323

324 *Explained variance and random forest classification of HMO and metagenome data*

325 Using the cross-sectional HMO and cross-sectional metagenome dataset, we sought to determine
326 which clinical factors were most associated with the HMO and the bacterial profiles (**figure 5a**).
327 Secretor status explained 56% of the variance within HMO profiles (adj. P < 0.001), but no other
328 co-variate was significantly associated with the HMO profiles. In contrast, the bacterial profiles
329 were significantly associated with both postmenstrual age (R² 0.07; adj. P = 0.006) and day of life
330 (R² 0.07; adj. P = 0.006), as well as receipt of antibiotics at the time of sampling (R² 0.06; adj. P =
331 0.006) and receipt of probiotics (R² 0.12; adj. P = 0.006), but not maternal secretor status (R² 0.02;
332 adj. P = 0.58). Together, these findings highlight that HMO and bacterial profiles are influenced by
333 numerous non-overlapping factors related to early life in preterm infants.

334 We compared the performance of random forest classification models built on the cross-sectional
335 subset of HMO profile data, metagenomic sequencing data, and the two datasets combined to
336 classify an infant as NEC or healthy, given that all this information is available before onset of
337 disease and could therefore function as a risk stratification system in clinical practice. The HMO
338 profile alone had a classification error of 0.146, with 21% (3/14) NEC and 12% (4/34) control
339 infants misclassified. DSLNT had the greatest contribution to classification with a Mean Decrease
340 Accuracy (MDA) of 0.11. Other HMOs contributing to classification accuracy included LNH

341 (MDA = 0.012) and DFLNH (MDA = 0.011), which were non-significantly higher in NEC infants.
342 Random forest generated using the metagenomic sequencing data was characterised by a
343 classification error of 0.229, with 43% (6/14) NEC and 15% (5/34) control infants misclassified.
344 *Enterobacter cloacae* was the most important feature guiding the classification (MDA = 0.036),
345 with higher relative abundance in NEC infants, followed by *Bifidobacterium bifidum* (MDA =
346 0.024) and *Bifidobacterium longum* (MDA = 0.013) which had higher relative abundance in control
347 infants. Combining HMO and metagenome datasets slightly improved the performance compared to
348 using HMOs alone, with 21% (3/14) NEC infants and (9%) 3/34 controls misclassified. In this
349 combined model, DSLNT was enriched in controls and DSLNH and the relative abundance of
350 *Escherichia unclassified* were higher in NEC infants (**figure 5b**).

351

352 **Discussion**

353 Receipt of human breast milk and early life gut microbiome development are intrinsically linked
354 and both influence the risk of NEC in preterm infants. Our study represents the largest analysis of
355 HMOs in NEC and the first to integrate HMO and metagenome data. We found DSLNT was
356 present in significantly lower concentrations in MOM fed to infants diagnosed with NEC.
357 Furthermore, lower DSLNT concentrations in MOM were associated with reduced transition into
358 PGCTs typically observed in older infants and lower relative abundance of *Bifidobacterium* spp.
359 The HMO results from the current study build upon previous findings in humans, showing reduced
360 DSLNT in MOM received by infants developing NEC, independent of maternal secretor status (17-
361 19). This is also supported by rodent studies where total and individual HMOs including 2'FL and
362 DSLNT have shown a protective effect against NEC development (16, 36, 37). However, 2'FL and
363 mixtures of HMOs (one of which included DSLNT) did not show any protection in NEC piglet
364 models (38, 39). Importantly from a clinical perspective, in rats the protection provided by pooled
365 HMOs could be reproduced with DSLNT alone, with specific dependence on its precise structure
366 since closely related sialyllacto-N-tetraose (identical in structure to DSLNT but lacking one sialic

367 acid residue) did not provide protection, suggesting a highly structure-specific mechanism (16). Our
368 findings further extend the evidence for the specificity of DSLNT in the NEC pathway. A threshold
369 level of DSLNT (241 nmol/mL) from a single MOM sample correctly identified 91% of NEC
370 infants (below threshold) and 86% of control healthy infants (above threshold). Of the three infants
371 who developed NEC despite a DSLNT above the threshold, two had not received MOM in the 3
372 weeks prior to disease onset and the remaining infant had a DSLNT concentration of 248 nmol/ml.
373 Within the validation dataset (17), 100% NEC infants were correctly classified, but only 60% of
374 controls. Making a robust diagnosis of NEC is difficult and it is possible that the specific threshold
375 value of DSLNT we identified will have a different predictive value in other populations or where
376 other criteria are used to determine the presence of disease. Our study contains a large number of
377 cases coded clinically as NEC independently validated by blinded review. In addition, our cohort
378 was more homogenous (predominantly white Caucasian) and the concentration of DSLNT less
379 variable (current study IQR 184-321 nmol/mL vs. Aufran *et al.* IQR 122-346 nmol/mL) despite
380 using the same analytical platform. Given HMO composition and DSLNT concentrations may be
381 influenced by genetic factors, geographical location, ethnicity (40), and seasonality (15), differential
382 thresholds may improve diagnostic performance in other settings. Taken together, this external
383 validation and potential variation in DSLNT concentration by maternal factors underscore the need
384 for large multicentre studies to both refine a universal or stratified threshold for DSLNT
385 concentration in predicting NEC and potentially prospectively identifying milk samples that may
386 benefit from supplementation with synthetically produced DSLNT.

387 In addition to HMO profiles, our extensive longitudinal stool metagenomic analysis, represents one
388 of the largest datasets to date. This extends our previous work (20, 33, 41) where DMM was used to
389 facilitate analysis of temporal microbiome development, and integrate the HMO DSLNT threshold
390 of 241 nmol/ml with infant gut microbiome profiles. We observed a difference in microbiome
391 development between DSLNT groups, with infants receiving MOM with lower DSLNT tending to
392 have delayed progression into the PGCT typically expected in older infants (i.e., PGCT-5). This

393 supports the theory that concentrations of specific HMOs in MOM are associated with differences
394 in gut microbiome development. On the contrary, transition into PGCT-5 was twice as likely in
395 infants receiving MOM with DSLNT above the threshold, which was characterised by high relative
396 abundance of *Bifidobacterium* spp. *Bifidobacterium* has previously been linked to health in preterm
397 infants (20, 41, 42) and our current findings in pre-NEC samples further support the association of
398 reduced *Bifidobacterium* spp., specifically *Bifidobacterium longum*, as a risk factor for NEC. In
399 addition, our species level metagenome data advanced previous associations of Enterobacteriaceae
400 with NEC (21, 23, 43), showing *E. cloacae* relative abundance was higher before NEC.

401 Random forest analysis confirmed the capability of HMO profiles to identify infants who developed
402 NEC and slightly outperformed metagenome profiles by correctly classifying three more NEC cases
403 and one more control. Combining HMO and metagenome data before disease accurately classified
404 87.5% of infants as healthy or NEC, with DSLNT and the bacterial species identified as important
405 in the random forest analysis being comparable to the unsupervised analysis in the current study and
406 in previous studies. Further work is needed to determine if DSLNT functions via modulation of the
407 microbiome or by acting directly on the host, such as acting in a structure-specific receptor-
408 mediated way to alter immune functioning and reduce inflammation leading to necrosis. In the
409 event of the latter, a microbial community with less DSLNT utilisation could provide an advantage
410 to reducing NEC risk. Taken together, the current findings and recent work highlighting the ability
411 of *Bifidobacterium* spp. to utilise HMOs is strain specific (7, 8) underscore the need for further
412 research to better understand the complexity of human milk and other nutritional exposures,
413 including the use of supplements such as prebiotics and probiotics in preterm infants. In addition to
414 therapeutics, the classifiers may provide a basis for the development of biomarkers predicting NEC
415 risk. While additional work is needed, the addition of microbial biomarkers may allow for the most
416 accurate predictions and could inform NEC risk for infants where MOM (and thus HMO
417 information) is not available.

418 This study involved the largest cohort to date investigating the relationship between HMO
419 composition and NEC development, and includes one of the most extensive longitudinal stool
420 metagenomic analyses of preterm infants. However, there are several limitations and avenues for
421 future work. First, the cross-sectional HMO profiling data precluded assessment of changes within
422 mothers over time and how this may relate to NEC development. The milk sample was selected
423 based on the day of infant feeding and the actual expression of milk may have occurred several days
424 earlier, which may be important clinically. Current published data suggest that the concentration of
425 HMOs, including DSLNT, are relatively stable over time (14), but validation in longitudinal
426 preterm cohorts is needed. Second, the amount of MOM an infant receives and tolerates each day is
427 variable, and DSLNT exposure is dependent on both concentration and volume. Although this study
428 identifies DSLNT concentration alone may be useful from both a diagnostic and therapeutic
429 perspective, further studies could consider the volume of milk received in addition to concentration.
430 Thirdly, inclusion of metagenome data was opportunistic based on available data and cost
431 prohibited sequencing all infants in the cohort. As such, the classification accuracy of the model
432 might be impacted by the reduced sample size in comparison to the full cohort, necessitating the
433 need for follow-up analyses in larger cohorts. Despite this, the sample size of 644 including 195
434 samples from 14 preterm infants who developed NEC makes this dataset one of the largest
435 published to date. Finally, the gene relative abundance data warrants further investigation, in
436 combination with other experimental approaches, to help inform the HMO utilisation capacity of
437 different strains.

438 In summary, HMO profiling of MOM coupled to metagenomic sequencing of preterm stool showed
439 that the concentration of a single HMO, DSLNT, was lower in milk received by infants who
440 developed NEC. The lower concentration of DSLNT was associated with altered microbiome
441 development, specifically a reduced progression toward the PGCT typically found in the older
442 infants which was abundant in *Bifidobacterium* spp. These results suggest MOM HMO profiling
443 may provide potential targets for biomarker development and disease risk stratification. They may

444 also guide focussed donor milk use (e.g., prioritise high DSLNT for preterm infants) and novel
445 avenues for supplements that may prevent life-threatening disease.
446

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544

545

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549

550 **Contributions**

551 NDE, JEB, and CJS conceived and designed the study. ACM, CAL, GY, CLG, JEB, and CJS
552 collected the samples and overseen the logistics. JAN and LB performed the HMO profiling. KLH,
553 and JFP performed the bioinformatics on fastq files. ACM, DPS, and CJS performed the analysis.

554 NDE, JEB, and CJS supervised the study. ACM, NDE, CAL, JEB, and CJS co-wrote the
555 manuscript and all authors approved the final submission.

556

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562

563 **Competing interest**

564 CJS declares performing consultancy for Astarte Medical and honoraria from Danone Early Life
565 Nutrition. NDE declares research funding from Prolacta Biosciences US and Danone Early life
566 Nutrition, and received lecture honoraria from Baxter and Nestle Nutrition Institute, but has no
567 share options or other conflicts. LB is UC San Diego Chair of Collaborative Human Milk Research,
568 endowed by the Family Larsson-Rosenquist Foundation, and serves on the foundation's scientific
569 advisory board. LB is (co-)inventor on patent applications regarding human milk oligosaccharides
570 in prevention of necrotizing enterocolitis and other inflammatory disorders. The other authors
571 declare that they have no competing interests.

572

573 **Ethical approval**

574 Ethics approval was obtained from the County Durham and Tees Valley Research Ethics
575 Committee (REC10/H0908/39) and parents gave informed consent for stool and data collection.

576

577 **Data availability statement**

578 Data are available upon reasonable request. All sequencing data generated and analysed in this
579 study have been deposited in the European Nucleotide Archive (ENA) under study accession
580 number PRJEB39610.
581

582 **Tables**

583 **Table 1. Demographics of the analytical cohort with human milk oligosaccharide profile data.**

584 Differences between groups were tested applying Chi-square test and Dunn’s post-hoc test where
 585 applicable.

	Control	NEC	P value
Number of patients	37	33	-
Secretors	25 (68%)	20 (61%)	0.544
Male	14 (38%)	22 (67%)	0.016
Vaginal delivery	25 (68%)	17 (52%)	0.171
Gestational age	25 [24; 26]	25 [24; 27]	0.881
Birthweight	670 [585; 830]	670 [600; 840]	1.000
Probiotics ever	37 (100%)	31 (94%)	0.855
MOM only	3 (8%)	6 (18%)	
MOM + Formula	11 (30%)	12 (37%)	
MOM + BMF	10 (27%)	7 (21%)	0.468
MOM + Formula + BMF	13 (35%)	8 (24%)	
DOL breast milk sample	18 [12; 31]	18 [13; 34]	0.636
DOL disease onset	-	19 [14; 35]	-
NEC surgical	-	16	-

586 NEC, necrotising enterocolitis; MOM, mother’s own breast milk; BMF, breast milk fortifier;

587 DOL, day of life

588

589

590

591 **Figure Legends**

592

593 **Figure 1. Analysis of human milk oligosaccharide (HMO) profiles and DSLNT concentration**

594 **in necrotising enterocolitis (NEC) and controls. a,** Orthogonal partial least squares discriminant

595 analysis (OPLS-DA) of maternal HMO profiles fed to infants diagnosed with NEC and controls.

596 The P value was calculated based on 2000 permutations. **b,** Visual representation of P values

597 obtained from comparison of individual HMOs between NEC and control group. Wilcoxon rank

598 sum test was applied, and P values adjusted with FDR algorithm. The line indicates P value = 0.05.

599 **c,** Univariate receiver operating characteristic (ROC) curve generated on DSLNT concentration

600 identified 241 nmol/ml as the best threshold for NEC prediction. The performance of the
601 classification is defined by the area under the curve, specificity (false positive rate) and sensitivity
602 (false negative rate). **d**, Box plot showing the concentration of disialyllacto-N-tetraose (DSLNT)
603 between NEC and controls. Blue line represents the 241 nmol/ml threshold.

604

605 **Figure 2. Analysis of human milk oligosaccharide (HMO) profiles with stratification of**
606 **necrotising enterocolitis (NEC) into medical (NEC-M) and surgical (NEC-S).** **a**, Partial least
607 squares discriminant analysis (PLS-DA) of HMO profiles from control, NEC-M, and NEC-S
608 infants. NEC-M and NEC-S cluster together and separately from controls ($P < 0.001$). P values
609 were calculated based on 2000 permutations. Box plots of **(b)** disialyllacto-N-tetraose (DSLNT) and
610 **(c)** lacto-N-neotetraose (LNnT) concentration between control, NEC-M, and NEC-S infants.
611 Kruskal-Wallis followed by Dunn's test using Bonferroni adjustment was applied. **d**, Adjusted
612 linear regression model for DSLNT and LNnT including potential clinical confounders. P values
613 were corrected by FDR. Significant variables are indicated by asterisks: *** denotes $FDR P <$
614 0.001 ; ** denotes $FDR P > 0.01$. DOL, day of life; PMA post-menstrual age; GA, gestational age.

615

616 **Figure 3. Cross-sectional analysis of preterm stool metagenome profiles between necrotising**
617 **enterocolitis (NEC) and matched controls.** Analysis includes the sample closest NEC onset
618 (median of 3 days prior to NEC) and a corresponding control sample matched by day of life. **a**,
619 Alpha-diversity based on observed species (richness) and Shannon diversity. **b**, Bray-Curtis
620 principal coordinate analysis. **c**, Box plots showing the relative abundance of significant phyla. **d**,
621 Box plots showing the relative abundance of significant species.

622

623 **Figure 4. Analysis of preterm gut community types (PGCTs) by infants receiving maternal**
624 **milk above or below the 241 nmol/mL DSLNT threshold.** The entire dataset of 644 samples
625 formed five distinct clusters based on lowest Laplace approximation following Dirichlet
626 multinomial clustering. **a**, Heatmap showing the relative abundance of dominant bacterial species
627 within each PGCT cluster. The phyla for each species are also shown. **b**, Transition model showing
628 the progression of samples through each PGCT, from day of life 0 to 60 across eight distinct time
629 points. Plots are separated based on whether the concentration of disialyllacto-N-tetraose (DSLNT)
630 in maternal milk was above or below the 241 nmol/mL threshold. Nodes and edges are sized based
631 on the total counts. Nodes are coloured according to DMM cluster number and edges are coloured
632 by the transition frequency. Transitions with less than 5% frequency are not shown.

633

634 **Figure 5. Modelling of cross-sectional human milk oligosaccharide (HMO) and infant stool**
635 **metagenomic profiles using Adonis and random forest. a,** horizontal bar plots showing the
636 variance (r^2) in maternal HMO and infant stool metagenomic profiles explained by clinical
637 covariates as modelled by univariate Adonis. Variables with an FDR $P < 0.05$ are shown in red.
638 DOL, day of life; PMA, post-menstrual age. **b,** Feature importance from combined HMO and
639 metagenome random forest classification model. Mean decrease accuracy (MDA) value defines the
640 contribution given by a certain feature to classification process.
641

642 **Online supplementary figure Legends**

643

644 **Online supplementary figure 1. Sampling schematic for the entire cohort.** Only samples
645 collected in the first 100 days of life are shown. Shapes represent sample timing in relation to the
646 diagnosis of disease; control (n = 37) and necrotising enterocolitis (NEC; n = 33). Colours indicate
647 if the sample on that day of life was a maternal breast milk human milk oligosaccharide (HMO) or
648 infants stool metagenome, or if both sample/data types were generated from each sample collected
649 on that day.

650

651 **Online supplementary figure 2. Comparison of human milk oligosaccharide (HMO) profiles**
652 **by maternal secretor status.** All infants were included (n = 77). **a**, Principal component analysis
653 (PCA) showing the clustering of HMO profiles based on secretor status. **b**, Visual representation of
654 P values for comparison of individual HMOs between secretor and non-secretor groups. 16 of the
655 19 HMOs were different between the two groups. Wilcoxon rank test was applied, and P values
656 were adjusted using FDR algorithm. **c**, HMO Shannon diversity was higher in breast milk from
657 secretor mothers compared to non-secretors. Wilcoxon rank test was applied. **d**, Stacked bar plot of
658 HMOs concentrations describing HMO profile of each breast milk sample analysed. 2'FL used for
659 identifying secretor status is almost absent in non-secretor breast milks, and present in relatively
660 high concentration in samples from secretor mothers.

661

662 **Online supplementary figure 3. Human milk oligosaccharide (HMO) profiles and**
663 **disialyllacto-N-tetraose (DSLNT) are different between necrotising enterocolitis (NEC) and**
664 **control group independent of maternal secretor status.** Orthogonal partial least squares
665 discriminant analysis (OPLS-DA) of breast milk HMO profiles from secretors (**a**) and non-secretors
666 (**b**). P values were calculated performing 2000 permutations. Comparison of DSLNT concentration
667 in milk received by NEC or controls separated by (**c**) secretor and (**d**) non secretor status. DSLNT
668 concentration is lower in milk received by the NEC group independently of secretor status. Group
669 comparison was performed applying Wilcoxon rank test and P values adjusted using FDR.

670

671

672 **Online supplementary figure 4. Shannon diversity of human milk oligosaccharides (HMOs)**
673 **was not associated with NEC development.** Shannon diversity of (**a**) overall cohort, (**b**) secretor
674 group, and (**c**) non-secretor group.

675

676

677 **Online supplementary figure 5. Human milk oligosaccharide (HMO) profiles were predictive**
678 **of necrotising enterocolitis (NEC) status. a,** Receiver operating characteristic (ROC) curves
679 generated using linear support vector machine (SVM) classification of HMO profiles between NEC
680 and control groups. Increasing numbers of HMO were included in the model and performance was
681 described by the AUC. Two HMOs model gave the optimal performance. **b,** Feature importance for
682 the two HMOs model. Disialyllacto-N-tetraose (DSLNT) was selected as discriminatory feature in
683 100% of the permutations. **c,** Box plot showing the concentration of disialyllacto-N-tetraose
684 (DSLNT) between NEC and controls from the validation dataset, Autran *et al.* (2018). Blue line
685 represents the 241 nmol/ml threshold.

686

687

688 **Online supplementary figure 6. Disialyllacto-N-tetraose (DSLNT) and lacto-N-neotetraose**
689 **(LNnT) concentrations were not influenced by the day of life (DOL) of sample.** Plot of (a)
690 DSLNT and (b) LNnT concentrations in relation to DOL of the sample. P values were calculated by
691 linear regression and DSLNT and LNnT concentrations were not related to the DOL the breast milk
692 was fed to the infant.

693