Investigation of the impact of stool collection methods on ¹ the metabolomics analysis/profiles of infant fecal samples ²

Chiara-Maria Homann ¹, Sara Dizzell ¹, Sandi M. Azab ², Eileen K. Hutton ^{3,4} on behalf of the GI-MDH Consortium⁺, Katherine M. Morrison ^{5,6} and Jennifer C. Stearns ^{1,7,*}

Department of Medicine, McMaster University, Hamilton, Ontario, Canada; chiara.m.homann@hotmail.com (CM.H.); sara.diz-	5
<u>zell@gmail.com</u> (S.D.); <u>surette@mcmaster.ca</u> (M.G.S.)	6
Department of Chemistry and Chemical Biology, McMaster University, Hamilton, Ontario, Canada; azabs@mcmaster.ca (S.M.A.)	7
Department of Obstetrics and Gynecology, McMaster University, Hamilton, Ontario, Canada; huttone@mcmaster.ca (E.K.H.)	8
McMaster Midwifery Research Centre, McMaster University, Hamilton, Ontario, Canada	9
Department of Pediatrics, McMaster University, Hamilton, Ontario, Canada; morriso@mcmaster.ca (K.M.M.)	10
Centre for Metabolism, Obesity and Diabetes Research, McMaster University, Hamilton, Ontario, Canada	11
Farncombe Family Digestive Health Research Institute, McMaster University, Hamilton, Ontario, Canada	12
The GI-MDH Consortium members are provided in the Acknowledgements.	13

* Correspondence: stearns@mcmaster.ca

Abstract: Metabolomic studies are important to understand microbial metabolism and interaction between the host and 15 the gut microbiome. Although there have been efforts to standardize sample processing in metabolomic studies, infant 16 samples are mostly disregarded. In birth cohort studies, the use of diaper liners is prevalent and its impact on fecal 17 metabolic profile remains untested. In this study, we compared metabolite profiles of fecal samples collected as solid stool 18 and those collected from stool saturated liner. One infant's stool sample was collected in triplicate for solid stool and stool 19 saturated liner. Comprehensive metabolomics analysis of the fecal samples was performed using NMR, UPLC and DI-MS. 20 The total number, identities and concentrations of the metabolites were determined and compared between stool sample 21 collection methods (stool vs. liner). The number and identity of metabolites did not differ between collection methods for 22 NMR and DI-MS when excluding metabolites with a coefficient of variation (CV) > 40%. NMR analysis demonstrated 23 lowest bias between collection methods, and lowest technical precision between triplicates of the same method followed 24 by DI-MS then UPLC. Concentrations of many metabolites from stool and stool saturated liner differed significantly as 25 revealed by Bland-Altman plots and t-tests. Overall, a mean bias of 10.2% in the Bland-Altman analysis was acceptable for 26 some metabolites confirming mutual agreement but not for others with a wide range of bias (-97-117%). Consequently, 27 stool and stool-saturated liner could be used interchangeably only for some select metabolite classes e.g. amino acids. 28 Differences between the metabolomic profiles of solid stool samples and stool saturated liner samples for some important 29 molecules e.g., ethanol, fumarate, short chain fatty acids and bile acids, indicate the need for standardization in stool 30 collection method for metabolomic studies performed in infants. 31

Keywords: stool sample collection; infant fecal metabolome; infant gut; stool; diaper liner

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1. Introduction

The microorganisms that reside within the human gut intimately interact with the host -36 immunologically, and metabolically [1]. Methods to study the gut microbiome include DNA and RNA based 37 methods that provide information about microbial genes and pathways; however, these methods can only 38 predict microbial and host metabolism. In order to study metabolic differences associated with health and 39 disease, microbial and host metabolites can be measured directly using metabolomics. Metabolomics is the 40 high-throughput identification and quantification of small molecules in body tissue or biofluids [2]. It is one 41 of the newer -omics technologies/disciplines [3]. Gut metabolites are very diverse and many of them remain 42 uncharacterized. As such, a number of different methods have emerged to resolve and quantify the 43 constituents and diversity of the gut metabolome. 44

The gut microbiota is often explored in relation to the fecal metabolite profile and there is high 45 interrelation between the gut microbiome and the fecal metabolome [4]. The fecal metabolome has been said 46 to provide a "complementary functional readout" of microbial metabolism as well as the interaction between 47 the gut microbiome and the host [4]. The relationship between metabolites and the microbiota within the 48 human gut has been explored in adults [5-7]; however, less is known about the gut metabolome in infants. 49 Establishment of the infant gut microbiome occurs in the first few years of life, which is a critical time in 50 development and is emerging as an important predictor of later health outcomes [8-10]. Currently, the most 51 common analysis of the gut microbiome in infants is performed using culture independent methods, such as 52 shotgun metagenomic sequencing or taxonomic profiling with use of the 16S rRNA marker gene. Standard 53 protocols are important for reproducibility in research and within the microbiome field some important work 54 has been done to document the variation that arises due to technical and methodological differences between 55 studies and study centers [11-14]. 56

Standard protocols exist to collect adult [15] and infant stool for nucleic acid analysis, but whether these 57 same collection methods are appropriate for metabolites is not known. There is a lack of standardization in 58 metabolomics study protocols and sample collection methods, especially for fecal metabolomics analysis [3]. 59 Additionally, studies that investigate methodological differences and their implications have solely been 60 performed using adult samples [3]. Infants produce stool of varying consistencies and breastfeeding babies in 61 particular often do not have solid stool, making sample collection challenging. One of the approaches to the 62 challenge of infant stool samples has been implementing the use of a standard diaper liner to collect stool 63 residue if solid stool is not available. Therefore, it is important to determine if similar metabolite profiles, in 64 terms of the type and number of unique metabolites, are obtained when different methods of sample collection 65 are used, namely from solid stool or stool saturated diaper liner. In this paper we compared these two 66 collection methods from one stool sample in triplicate across three metabolomic platforms (nuclear magnetic 67 resonance, NMR; direct infusion mass spectrometry, DI-MS, and ultra-high performance liquid 68 chromatography, UPLC) to determine the metabolic and technical variation introduced by the collection 69 method. 70

2. Results

2.1. Metabolite Coverage

After removing metabolites with coefficient of variation (CV) > 40%, 159 unique metabolites were 74 detected altogether in the solid stool and stool saturated liner samples (Figure 1A). 65 metabolites were 75 detected by NMR, 79 by DI-MS and 39 by UPLC. While none of the metabolites detected by UPLC were 76 measured with the other platforms, 24 metabolites were measured by both NMR and DI-MS (Figure 1A). The 77 overlapping metabolites included mostly amino acids e.g., alanine, asparagine, isoleucine, leucine, and valine., 78 saccharides e.g., glucose, and short-chain fatty acids e.g., acetate, butyrate, and propionate. To assess the 79 variation of metabolites within a sample for the two sample collection methods, the chemodiversity index was 80 calculated for each sample based on the number of unique metabolites across all three metabolomic methods. 81 The chemodiversity index did not differ significantly between sample collection methods (p = 0.32), which 82 indicates that the same metabolites are found in solid stool samples and stool saturated liners. This is further 83 confirmed by the number of metabolites found with each sample collection method and metabolic platform, 84 where metabolite overlap between stool collection methods ranges from 95.0 - 100% (Table 1, Figure 1A), prior 85 to and after the exclusion of metabolites with CV > 40%. A Bland-Altman plot was generated to assess overall 86 mutual agreement between sample collection methods (Figure 1B). Although, the mean bias between stool 87 collection methods was 10.2% (p = 0.019) for 159 metabolites, the 95% confidence interval was wide, indicating 88 that concentrations differed by stool collection method and suggesting that stool saturated liner samples had 89

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average metabolite concentrations that were lower than average metabolite concentrations in solid stool90samples. Additionally, a few metabolites were outside of the limits of agreement for all metabolomic platforms91(indicated by arrows in Figure 1B), signifying that there were drastic differences in the concentrations for some92metabolites e.g., ethanol, formate, and fumarate, creatinine and arginine based on stool sample collection93method.94



Figure 1. (**A**) Number of metabolites measured per platform (UPLC, NMR, DI-MS) after metabolites with a coefficient of variance greater than 40% were excluded; (**B**) Bland-Altman plot comparing metabolite concentrations between methods for stool collection (solid stool vs. liner), based on sample rank and colored by metabolic platform. Metabolites outside of the confidence intervals of the upper and lower limits are considered different between methods and are indicated by the black arrows.

	Number of metabolites measured (n)	Number of metabolites detected (n)	Number of me- tabolites de- tected in solid stool (n, %)	Number of me- tabolites de- tected in liner (n, %)	Overlap be- tween solid stool and liner (n, %)		
No CV cut-off							
NMR	67	67	67 (100)	67 (100)	67 (100)		
DI-MS	116	110	109 (99.1)	108 (98.2)	107 (97.3)		
UPLC	68	40	40 (100)	38 (95.0)	38 (95.0)		
CV < 40%							
NMR	67	65	65 (100)	65 (100)	65 (100)		
DI-MS	116	79	79 (100)	79 (100)	79 (100)		
UPLC	68	39	39 (100)	37 (95.0)	37 (95.0)		

 Table 1. Comparison of the number of metabolites measured using different metabolomic methods and collection methods.

2.2. Metabolites Measured with Nuclear Magnetic Resonance (NMR)

Quantitative NMR spectroscopy was used for targeted metabolomic analysis of water-soluble metabolite 106 classes including amino acids, saccharides, alcohols, organic acids, amines, tricarboxylic acid (TCA) cycle 107 intermediates, and short chain fatty acids (SCFAs). A total of 67 metabolites were measured and 65 were 108 detected in both solid stool and stool saturated liner samples (CV < 40%) (Table 1). Ethanol, formate, and 109 fumarate were outside of the limits of agreement of the Bland-Altman plot (Figure 1B), indicating significant 110 differences between sample collection methods. Ethanol and fumarate had higher concentrations in solid stool 111 samples, whereas creatinine had higher concentrations in stool saturated liner samples. In further analyses 112 with paired t-tests, the concentrations of 56 metabolites were significantly different between stool saturated 113 liner and solid stool samples after adjustment for multiple testing (Figure S2A). Of the physicochemical 114 characteristics, only polar surface area was significantly associated with the percentage difference in 115 concentrations between the stool sample collection methods. Higher polar surface area was associated with 116 higher metabolite concentrations in stool saturated liner samples ($\beta = -0.17$, p = 0.017). Metabolite 117

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characteristics such as polarizability, solubility in water, molecular weight, or physiological charge were not 118 associated with the difference in concentration between solid stool and stool saturated liner (Figure S2B, p >119 0.05). Variation between technical replicates was lower for metabolite measurements by NMR compared with 120 the variation seen for the other metabolic platforms. Mean (SD) variation for solid stool samples was 2.8% (SD 121 = 2.50), and average variation for stool saturated liner samples was 3.4% (SD = 2.87) (Table 2). 122

Table 2. Coefficients of variation by metabolic platform after metabolites with coefficients of variations > 40% were excluded.

Solid Stool					Diaper Li	ner
	Mean (SD)	Median	Range (Min, Max)	Mean (SD)	Median	Range (Min, Max)
NMR	2.81 (2.50)	2.22	12.32 (0.059, 12.38)	3.37 (2.87)	2.57	16.19 (0.19, 16.39)
DI-MS	10.03 (9.62)	8.13	34.54 (0.00, 34.60)	8.75 (8.79)	7.51	34.64 (0.0, 34.64)
UPLC	15.48 (6.66)	14.43	31.36 (6.19, 37.55)	17.06 (5.52)	16.58	26.85 (5.88, 32.73)

2.3. Metabolites Measured with Direct Flow Injection Mass Spectrometry (DI-MS)

For targeted metabolomic analysis of biogenic amines, amino acids, acylcarnitines, phospholipids and 128 sphingolipids, direct flow injection mass spectrometry (DI-MS) was used. A total of 116 metabolites were 129 measured and 79 of these metabolites were detected in both solid stool and stool saturated liner samples after 130 the CV cut-off (Table 1). For this metabolic platform, only one of the metabolites, arginine, was outside of the 131 limits of agreement in the Bland-Altman plot (Figure 1B). Significant differences were found in the 132 concentration of DI-MS measured metabolites in stool saturated liner and solid stool samples in paired t-tests 133 after adjustment for multiple testing for 16 metabolites (Figure S3A). There was noticeable technical variation 134 with this method, however, which reduced precision of these measurements. Mean (SD) CV for solid stool 135 samples was 10.0% (SD = 9.62) and 8.8% (SD = 8.79) for stool saturated liner samples (Table 2). None of the 136 metabolite characteristics were associated with the difference in metabolite concentrations between solid stool 137 and stool saturated liner samples (Figure S3B, p > 0.05).

2.4. Metabolites Measured with Ultra-High Performance Liquid Chromatography (UPLC)

UPLC was used as a targeted metabolomic method to examine bile acids in the stool samples. A total of 141 68 metabolites were measured, however, only 39 metabolites were retained after the CV cut-off; all 39 were 142 detected in solid stool and 37 metabolites were detected in stool saturated liner (Table 1), 143 Glycochenodeoxycholic acid and isolithocholic acid were not detected in stool saturated liner. No metabolite 144 had a significantly higher or lower concentration within solid stool versus stool saturated liner samples in 145 paired t-tests after adjustment for multiple testing (Figure S4A). As expected, the concentration of the 146 metabolites measured with UPLC was ten-fold lower than with the other methods, with many metabolites 147 recorded near the limit of detection due to their low abundance, and there was more technical variation than 148 with NMR which may have reduced precision. Mean (SD) of CV for solid stool samples was 15.5% (6.66) and 149 17.1% (5.52) for stool saturated liner samples (Table 2), which were the highest average CVs of the three 150 metabolic platforms. As previously observed for DI-MS, metabolite characteristics were not associated with 151 their concentration measured from solid stool or stool saturated liner samples (Figure S4B, p > 0.05). 152

2.5. Short-Chain Fatty Acids

One of the metabolite groups of higher interest were the SCFAs, as these are known intermediates and 155 end-products of bacterial metabolism. Paired t-tests were performed to analyze the differences in absolute 156 concentrations between sample collection methods. The concentrations of all the SCFAs, as well as their total 157 concentration differed significantly by sample collection method. Acetate (p = 0.011), butyrate (p = 0.00084), 158 propionate (p = 0.0016), isovalerate (p = 0.0027), valerate (p = 0.00025) and total SCFAs (p = 0.0066) had higher 159 concentrations in the stool saturated liner samples, while isobutyrate (p = 0.0011) had a higher concentration 160 in the solid stool sample (Table 3). 161

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	Solid Stool			Diaper Liner			
	Mean (SD)	Median	Range (Min, Max)	Mean (SD)	Median	Range (Min,	Sig. ¹
						Max)	
Acetate	51.0 (2.24)	50.5	4.4 (49.0, 53.4)	59.6 (0.67)	59.7	1.3 (58.9, 60.2)	0.011
Butyrate	0.16 (0.0065)	0.16	0.02 (0.15, 0.17)	0.29	0.30	0.02 (0.28, 0.30)	0.00084
				(0.0108)			
Dromionato	19.8 (0.409)	19.9	0.8 (19.3, 20.1)	25.2	25.2	0.1 (25.2, 25.3)	0.0016
Propionate				(0.0905)			
Isovalerate	0.49 (0.0083)	0.49	0.02 (0.48, 0.50)	0.65	0.65	0.02 (0.64, 0.66)	0.0027
				(0.0070)			
Valorato	0.20 (0.0055)	0.20	0.01 (0.19, 0.20)	0.37	0.37	0.01 (0.36, 0.37)	0.00025
valerate				(0.0057)			
Icobuturato	1.8 (0.056)	1.8	0.11 (1.76, 1.87)	0.98	0.97	0.02 (0.97, 0.99)	0.0011
Isobutyrate				(0.0091)			
Total SCFA	73.4 (2.52)	72.4	4.8 (71.5, 76.3)	87.1 (0.61)	87.1	1.2 (86.5, 87.7)	0.0066
1° bile acids	339.2 (48.5)	342.3	96.9 (289.2, 386.1)	175.7 (36.1)	194.5	64.4 (134.1,	0.043
						198.5)	
2° bile acids	6.87 (0.75)	7.11	1.43 (6.03, 7.46)	4.58 (0.88)	4.70	1.74 (3.65, 5.39)	0.067*
Total bile ac-	346.1 (49.3)	349.4	98.4 (295.2, 393.6)	180.3 (36.9)	199.2	66.1 (137.8,	0.043
ids						203.9)	

Table 3. Concentrations of the metabolites of interest: SCFAs (µmol/g) and bile acids (nmol/g).

¹ Metabolite concentration differences between sample collection methods were tested using paired ttests.

2.6. Bile Acids

Another metabolite group of interest were the 15 major bile acids, as microbes are responsible for the 169 conversion of primary bile acids to secondary bile acids. These are cholic acid, chenodeoxycholic acid, 170 taurocholic acid, taurochenodeoxycholic acid, glycocholic acid, glycochenodeoxycholic acid, lithocholic acid, 171 deoxycholic acid, ursodeoxycholic acid, glycolithocholic acid, taurolithocholic acid, glycodeoxycholic acid, 172 taurodeoxycholic acid, glycoursodeoxycholic acid and tauroursodeoxycholic acid. Total concentrations for the 173 primary, secondary, and total bile acids were calculated and compared between sample collection methods 174 with paired t-tests. Primary bile acid concentrations and total bile acid concentrations were significantly higher 175 in solid stool samples than in stool saturated liner samples (p = 0.043; p = 0.043). Secondary bile acids were 176 trending towards a significantly higher concentration in solid stool samples (p = 0.067) (Table 3). 177

3. Discussion

Collection and banking of stool samples from current or previous cohorts for later metabolomic analysis 179 is an important activity. It is important to strike a balance between ease of sample collection and the 180 comparability of results across collection methods, especially in breastfeeding infants, where stool consistency 181 can provide a challenge for sample collection. In our study, Baby, Food & Mi, we have several infants for 182 which only stool saturated diaper liner was available. Undertaking an extensive study of fecal metabolites 183 measured using three different analytical techniques (NMR, DI-MS and UPLC) described here, we sought to 184 explore whether metabolomic profiles of solid stool samples are comparable to those found with stool 185 saturated liner. In this comparative study, we showed that after a cut-off for high technical variation, the 186 individual metabolites detected in solid stool samples and stool saturated liner did not differ between sample 187 collection methods; however, metabolite concentrations differed significantly between collection methods 188 when analyzed on all three metabolomic platforms: NMR, DI-MS and UPLC. Discrepancies were also seen for 189 metabolite groups of specific interest, namely SCFAs and bile acids. These metabolite groups in particular 190 showed differences in concentrations between solid stool and stool saturated liners, where all SCFAs except 191 isobutyrate, had higher concentrations in stool saturated liner samples and bile acids had higher 192 concentrations in solid stool samples, indicating that the sample collection method is an important 193 consideration in infant metabolomic studies. 194

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Metabolomic analysis performed in this study was based on two analytical systems: nuclear magnetic 195 resonance and mass spectrometry. NMR is a well-established platform [16] and results are highly 196 reproducible, inherently quantitative, robust, and cost effective [3,17,18]. However, NMR is less sensitive than 197 mass spectrometry, by a factor of 10 - 100x, and therefore, has a narrower coverage of metabolites [3,18]. 198 Reduced technical variation between samples from each collection method could surely impact the statistical 199 power of the t-tests performed and the resulting findings across all analytical platforms with an advantage 200 towards NMR that exhibits highest precision. As a high-throughput method, the advent of mass spectrometry 201 introduced a new dimension to medical research. Two mass spectrometry methods were described in this 202 study: UPLC-MS/MS and DI-MS. UPLC is a separation-based method, while DI-MS is separation free [19]. MS 203 methods are more sensitive than NMR and are often targeted for specific metabolites [17]. However, MS 204 methods have less reliable molecule quantification, need internal standards and are more prone to error due 205 to matrix effects [3,17]. Though not performed in this study, there is a high variety of MS-based methods for 206 metabolomic analysis including GC-MS, CE-MS and MALDI-MS; many studies primarily use NMR and GC-207 MS [19]. 208

Benchmarking the effect of the collection method (solid stool vs. stool saturated liner) on the resulting 209 metabolite concentrations is essential prior to undertaking large-scale analyses. This is especially important in 210 large birth cohort studies with breastfeeding infants, where the use of diaper liners is prevalent, and stool 211 often does not have a solid consistency. Factors affecting the metabolite profile in fecal samples are sample 212 collection methods, sample storage, as well as sample preparation [3]. Previous studies have indicated that 213 freezing does not affect the metabolic profile of stool samples. The suggested workflow for fecal metabolomic 214 samples is to keep fresh samples on ice until they reach the laboratory and can be processed. After initial 215 processing samples should be kept at -80 degrees Celsius before chemical analysis starts [2]. Multiple freeze-216 thaw cycles should be avoided since it could alter stool metabolite profiles [20]. The benefit of freezing samples 217 is that preservatives are not needed [15]. Multiple studies have investigated the use of preservatives in fecal 218 samples, assessing stability of the stool sample at room temperature. These studies have shown that 219 preservation in 95% ethanol shows the highest concordance with samples frozen quickly after collection - also 220 considered the gold standard for fecal sample collection [15,21]. In our study, however, stool sample 221 preservatives were not used, as sample was frozen quickly after defecation, following the evidence-backed 222 protocol outlined above. Other factors influencing fecal samples include stool water content, which was not 223 considered here. 224

Sample collection is an important consideration in metabolomics. To our knowledge, this study is the first 225 of its kind investigating differences between method of stool sample collection (solid stool vs. stool saturated 226 liner) in infant samples, and its impact on the metabolomics profiles. Therefore, this study provides valuable 227 insight into differences in technique for analyzing the metabolites of infant fecal samples. Limitations of this 228 study include the limited sample size (n = 1) and the high technical variation for some metabolites within the 229 three technical replicates for each method, especially for DI-MS and UPLC. Both of these factors reduce the 230 statistical power of the study, which might conceal actual associations when testing biological hypotheses. 231 Thus, to fully understand the methodological differences between solid stool and diaper liner samples this 232 study could be expanded to include more infants. After observing differences in metabolite concentrations for 233 all three of the metabolomic methods, NMR, DI-MS and UPLC, the practical implications of this study are to 234 solely use one of the sample collection methods for metabolomic analysis, preferably solid stool samples, as 235 stool saturated liner failed to detect one of the major bile acids glycochenodeoxycholic acid. Within a targeted 236 approach for studies only interested in the analysis of amino acids, solid stool and stool saturated liner could 237 be used interchangeably. 238

4. Materials and Methods

4.1. Participant Recruitment

One infant from the Baby, Food & Mi study [22], a sub-cohort of 15 infants from the primary Baby & Mi study at McMaster University in Hamilton, Ontario, Canada was observed [23]. 243

4.2. Fecal Sample Collection

The study participant was asked to use a diaper liner (Bummis, Quebec, Canada) during the sample 245 collection period; samples were collected at around 6 months of age. Fecal samples were collected by placing 246 the infant's soiled diaper, including the diaper liner in a resealable bag with an anaerobic sachet (Fisher 247

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Scientific, Hampshire, UK) immediately after the infant defecated. The bag containing the sample was then 248 placed in an insulated cooler bag with a frozen reusable ice pack and transported to McMaster University. 249 Processing of the sample occurred in a Bactron IV anaerobic chamber (Sheldon Manufacturing INC, Cornelius, 250 OR). Solid stool aliquots of 100 mg, 50 mg, 100 mg, and 600 mg were measured and aliquoted into cryovials 251 for DNA isolation, Direct Flow Injection Mass Spectrometry (DI-MS), Quantitative Nuclear Magnetic 252 Resonance Spectroscopy (NMR) and ultra-performance liquid chromatography-tandem mass spectrometry 253 (UPLC), respectively. Stool saturated liner samples containing 100 mg, 50 mg, 100 mg, and 600 mg of stool 254 were cut from the liner and also aliquoted into cryovials. To ensure that the amount of solid stool in the stool 255 saturated liner samples was accurate, a clean liner sample with no stool was weighed and subtracted from the 256 weight of the stool saturated liner. Clean liner was used as a negative control for metabolite extraction. All 257 samples (solid stool, stool saturated liner and clean liner) were aliquoted in triplicate for each metabolomic 258 method and were stored at -80 °C until further processing could occur. Samples were shipped on dry ice to 259 The Metabolomics Innovation Centre (TMIC; Alberta, Canada) where metabolic profiling was done according 260 to standard protocols, described briefly below. 261

4.3. Metabolomic Profiling

Samples were prepared according to [24] and [25]. Total metabolites were measured with nuclear 264 resonance spectrometry (NMR). All ¹H-NMR spectra were collected on a 700 MHz Avance III (Bruker) 265 spectrometer equipped with a 5 mm HCN Z-gradient pulsed-field gradient (PFG) cryoprobe. 1H-NMR spectra 266 were acquired at 25°C using the first transient of the NOESY presaturation pulse sequence (noesy1dpr), chosen 267 for its high degree of quantitative accuracy. All free induction decays were zero-filled to 250 K data points. 268 The singlet produced by the DSS methyl groups was used as an internal standard for chemical shift referencing 269 (set to 0 ppm). All 1H-NMR spectra were processed and analyzed using the Chenomx NMR Suite Professional 270 software package version 8.1 (Chenomx Inc., Edmonton, AB). Untargeted metabolomics was performed with 271 direct flow injection mass spectrometry with an Agilent 1100 series HPLC system (Agilent, Palo Alto, CA) and 272 an Agilent reversed-phase Zorbax Eclipse XDB C18 column (3.0 mm × 100 mm, 3.5 µm particle size, 80 Å pore 273 size) with an AB SCIEX QTRAP® 4000 mass spectrometer (AB SCIEX, CA, U.S.A.). The controlling software 274 was Analyst® 1.6.2. The mass spectrometer was set to positive electrospray ionization with multiple reaction 275 monitoring (MRM) mode. Bile acids were measured with Ultra-Performance Liquid Chromatography-276 Tandem Mass Spectrometry (UPLC) on an Agilent 1290 system coupled to a 4000 QTRAP mass spectrometer. 277 The MS instrument was operated in the multiple-reaction monitoring (MRM) mode with negative-ion (-) 278 detection. A Waters BEH 15-cm long, 2.1-mm I.D. and C18 LC column was used, and the mobile phase was 279 (A) 0.01% formic acid in water and (B) 0.01% formic acid in acetonitrile for binary-solvent gradient elution by 280 RPLC. Linear regression calibration curves were constructed between analyte-to-internal standard peak area 281 ratios (As/Ai) versus molar concentrations (nmol/mL). 282

4.4. Statistical Analysis

Data analysis was performed in R [26] and MedCalc [27]. Metabolites with coefficient of variation (CV) 285 greater than 40% were excluded from statistical analysis. MedCalc software was used to generate a Bland-286 Altman plot, comparing concentrations of the metabolites between solid stool and stool saturated liner for the 287 metabolomic platforms. For metabolites that were measured with two metabolomic platforms, the 288 metabolomic platform with the lower CV for the metabolite was included in the Bland-Altman plot. Statistical 289 significance between solid stool and stool saturated liner samples was determined with two-tailed paired t-290 tests with FDR-adjustment for multiple testing. Physical and chemical properties for each metabolite were 291 obtained from the Human Metabolome Database [28] and included solubility in water, physiological charge, 292 polarizability, polar surface area and molecular weight. Associations of metabolite characteristics with the 293 percentage difference in concentration between solid stool and stool saturated liner were tested using 294 univariate linear regressions with the stats package [26]. The chemodiversity index is an alpha diversity 295 measure that describes the variation of metabolites within a sample [29,30] and was calculated for each sample: 296

> Chemodiversity Index = 1- Number of metabolites with zero values / Total number of metabolites.

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(1)

Differences in chemodiversity index by sample collection method were calculated with a two-tailed paired ttest. For the investigation of specific metabolite groups (SCFAs and bile acids), individual and total SCFA concentrations, as well as primary, secondary, and total bile acid concentrations were calculated and compared between sample collection methods using two-tailed paired t-tests. The cut-off point for all statistical analyses presented here is p < 0.05.

5. Conclusions

This study establishes that there is an association between stool sample collection method and metabolite 308 profiles in three common metabolomic analysis methods, namely nuclear magnetic resonance, direct infusion 309 mass spectrometry and ultra-high performance liquid chromatography. This highlights the need to either 310 standardize research protocols to one of the stool collection methods or to control for stool collection method 311 in analyses. 312

Supplementary Materials: The following are available online at www.mdpi.com/xxx/s1, Table S1: CV for metabolites 313 measured in stool and liner with NMR, Table S2: CV for metabolites measured in stool and liner with DI-MS/MS, Table 314 S3: CV for metabolites measured in stool and liner with UPLC-MS, Figure S1: Bland-Altman plot comparing metabolite 315 concentrations between methods for stool collection (solid stool vs. liner), colored by metabolic platform. Metabolites 316 outside of the confidence intervals of the upper and lower limits are considered different between methods, Figure S2: 317 Metabolite, including short chain fatty acid, concentrations (µmol/g) measured in stool and liner with NMR. (A) Boxplot 318 of metabolite concentrations, * = significant differences between stool sample collection methods, after adjustment for 319 multiple testing. (B) Characteristics of the metabolites in relation to their log2 fold change by sample collection method., 320 Figure S3: Metabolite concentrations (µmol/g) measured in stool and from liner with DI-MS/MS. (A) Boxplot of metabolite 321 concentrations, * = significant differences between stool sample collection methods, after adjustment for multiple testing. 322 (B) Characteristics of the metabolites in relation to their log2 fold change by sample collection method., Figure S4: Bile acid 323 concentration (nmol/g) measured in stool and from liner with UPLC-MS. (A) Boxplot of metabolite concentrations. (B) 324 Characteristics of the metabolites in relation to their log2 fold change by sample collection method. 325

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- School of Nutrition and Translational Research in Metabolism (NUTRIM), Department of Medical Microbiology, 351
 Maastricht University Medical Centre, Maastricht, the Netherlands. 352
- ² in Vivo Planetary Health: an affiliate of the World Universities Network (WUN), West New York, New Jersey, USA. 353

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3	School for Public Health and Primary Care (CAPHRI), Department of Medical Microbiology, Maastricht University Medical Centre, Maastricht, the Netherlands.	354 355
4	Department of Epidemiology, Care and Public Health Research Institute (CAPHRI), Maastricht University, Maastricht, The Netherlands.	356 357
5	Department of Obstetrics and Gynecology, McMaster University, Hamilton, ON, Canada.	358
6	McMaster Midwifery Research Centre, McMaster University, Hamilton, ON, Canada.	359
7	Department of Pediatrics, McMaster University, Hamilton, ON, Canada.	360
8	Farncombe Family Digestive Health Research Institute, McMaster University, Hamilton, Canada	361
9	Department of Biochemistry & Biomedical Sciences, McMaster University, Hamilton, Canada	362
10	Department of Clinical Epidemiology & Biostatistics, McMaster University, Hamilton, Canada.	363
11	Department of Pediatric Pulmonology, Immunology and Intensive Care Medicine, Charité Universitätsmedizin Berlin, Germany.	364 365
12	Children's Center Bethel, Protestant Hospital Bethel, University of Bielefeld, Germany.	366
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1 2 3	Department of Obstetrics and Gynecology, McMaster University, Hamilton, ON, Canada. McMaster Midwifery Research Centre, McMaster University, Hamilton, ON, Canada. Department of Pediatrics, McMaster University, Hamilton, ON, Canada.	369 370 371
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