

# A longitudinal study of the faecal microbiome and metabolome of periparturient mares

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## ABSTRACT

**Background:** Periparturient mares are at increased risk of colic including large colon volvulus, which has a high mortality rate. Alterations in colonic microbiota related to either physiological or management changes, or both, that occur at this time have been suggested as potential causes for increased colic risk in this population of horses. Although the effect of management changes on the horse faecal microbiota has been investigated, limited work has been conducted to investigate changes in faecal microbiota structure and function in the periparturient period.

The objectives of the current study were to investigate temporal stability of the faecal microbiota and volatile organic compounds (VOCs) of the faecal metabolome in periparturient mares.

**Methods:** Faecal samples were collected weekly from five pregnant mares from 3 weeks pre-foaling to 7 weeks post-foaling. The microbiome data was generated by PCR amplification and sequencing of the V1–V2 regions of the bacterial 16S rRNA genes, while the VOC profile was characterised using headspace solid phase microextraction gas chromatography mass spectrometry.

**Results:** The mare faecal microbiota was relatively stable over the periparturient period and most variation was associated with individual mares. A small number of operational taxonomic units were found to be significantly differentially abundant between samples collected before and after foaling. A total of 98 VOCs were identified. The total number of VOCs did not vary significantly between individual mares, weeks of sample collection and feeds available to the mares. Three VOCs (decane, 2-pentylfuran, and oct-2-ene) showed significant increase overtime on linear mixed effects modelling analysis. These results suggest that the mare faecal microbiota is structurally and functionally stable during the periparturient period.

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Additional Information and  
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The findings also suggest that if changes in the gut microbiota are related to development of colic postpartum, altered risk may be due to inherent differences between individual mares. VOCs offer a cost-effective means of looking at the functional changes in the microbiome and warrant further investigation in mares at risk of colic.

**Subjects** Ecology, Microbiology, Veterinary Medicine

**Keywords** Horse, Mare, Colic, Microbiota, Microbiome, Metabolome, VOCs

## INTRODUCTION

Parturition and the periparturient period have been shown to be associated with increased risk of colic in several studies (*Kaneene et al., 1997*). Postpartum colic occurred in 11% of broodmares in a longitudinal study (*Weese et al., 2015*) and 36% of these cases were due to large colon volvulus (LCV). LCV is a form of colic that is associated with high mortality (*Driscoll et al., 2008; Harrison, 1988*) and increased risk of post-operative colic (*French et al., 2002*). Broodmares were 13 times more likely to develop LCV in one case control study (*Suthers et al., 2013*) and represented 74–77% of horses undergoing surgical management of LCV in two case series (*Ellis et al., 2008; Snyder et al., 1989*). Furthermore, around one-fifth of colic surgeries performed over a 24-year-period in a US equine referral hospital were for the correction of LCV in Thoroughbred mares (*Hackett et al., 2014*). Broodmares appear to be at greatest risk during the early postpartum period based on the fact that approximately 30% of horses diagnosed with LCV in two studies were postparturient mares that had foaled within 90 days prior to hospital admission (*Snyder et al., 1989; Suthers et al., 2013*).

The periparturient period is characterised by frequent changes in management including feeding practices, which have been consistently identified as risk factors for colic in general (*Cohen, Gibbs & Woods, 1999; Cohen et al., 1995; Cohen & Peloso, 1996; Hudson et al., 2001*) and for LCV (*Suthers et al., 2013*). It has been hypothesised that dietary and other management changes may disrupt colonic microbes and may induce changes in colonic pH and volatile fatty acid production, predisposing horses to intestinal dysfunction (*Cohen, Gibbs & Woods, 1999; Hudson et al., 2001*). Although perturbation of the horse gut microbial communities (known as gut microbiota) as a result of diet change is well-documented in microbiological studies (*Daly et al., 2012; Fernandes et al., 2014*), the effect of foaling or any associated hormonal or metabolic changes that occur during the periparturient period in these microbial communities is largely unknown.

Relatively little research has been conducted to determine if there are changes in the equine gut microbiota associated with foaling. Given the association between recent foaling and colic, and LCV in particular, this would provide valuable information about the potential role of gut microbiota in colic development. Faecal samples collected from broodmares at four time points (14 days prior to foaling and 4, 14, and 28 days post-foaling) in one study revealed stability of the faecal microbiota in this group of mares compared with non-pregnant control mares and when pre- and postpartum samples were

compared ([Weese et al., 2015](#)). However, the risk period for postpartum colic in broodmares appears to be greatest in the first 90 days post-foaling and the latter study did not investigate this longer period.

Volatile organic compounds (VOCs) are the products of metabolism of the microbiota and the host (mare). These compounds provide information about the functional microbiota (i.e. what the bacteria produce), which may be more important than identification of bacteria alone. In humans and animals, altered VOC profiles were shown to be non-invasive indicators of gastrointestinal disease ([Garner et al., 2007](#); [Leng et al., 2018](#)). Therefore, faecal VOC analysis may eventually provide a monitoring tool for types of equine colic that may arise from gut dysbiosis. The aim of the current study was to characterise the faecal microbiota and the faecal volatile metabolome of periparturient mares from 3 weeks pre-foaling to 7 weeks post-foaling, and to explore temporal stability of structure of the microbial community and the VOCs it produces.

## MATERIALS AND METHODS

### Mares

Five healthy pregnant mares from the same farm were recruited onto the study approximately 1 month prior to their foaling due dates. The mares had no history of colic or other medical conditions during pregnancy. The demographics of the mares are summarised in [Table S1](#). The mares' diet was supplemented with a feed balancer (Opti-Care Balancer; Gain Equine Nutrition, Durham, UK) throughout the study and they had free access to water from automatic watering devices. Pre-foaling and during the first few weeks post-foaling the mares were managed on grass paddocks during the day and stabled at night in separate foaling boxes where they were fed hay. Mares were then turned out in groups in larger grass paddocks with free access to lush grass until the end of the study. The timing of introduction of each of these nutritional management practices varied between mares and details are given in [Fig. S1](#). Moxidectin (Equest, Zoetis, Surrey, UK) was administered to all mares approximately 2 months prior to collection of the first set of samples, which was part of the normal management routine of the mares. The study was approved by The University of Liverpool Veterinary Research Ethics Committee (VREC207) and the manager of the farm consented to participate.

### Sample collection

A total of 11 samples (approximately 200 g each) were collected weekly from faeces passed by individual mares immediately following observed defaecation (samples labelled T-3 to T7). Samples were placed in plastic sealable bags and were stored in a refrigerator (4 °C) at the farm temporarily until all samples had been collected from mares scheduled for sampling on that day. Samples were then transferred to a -80 °C freezer where they were stored until processing. All samples were frozen within 5 h of collection except samples collected during the second stage labour (T0) where the yard staff collected these samples and refrigerated them until the next visit by the principal investigator. The sampling schedule and types of feed at each sampling occasion are depicted in [Fig. S1](#).

## DNA extraction and generation of sequence data

A commercial kit (QIAamp DNA Stool Mini Kit; QIAGEN, Manchester, UK) was used to extract DNA from samples followed by PCR amplification of the V1–V2 hypervariable regions of the bacterial 16S ribosomal ribonucleic acid (rRNA) gene to create amplicon libraries for sequencing using the Ion Torrent Personal Genome Machine system. Details of sample preparation and generation of sequence data are described previously (Salem *et al.*, 2018).

## Faecal volatile organic compound profiling

The VOC profile of faeces was determined using headspace solid phase microextraction gas chromatography mass spectrometry (HS-SPME-GCMS) using the methods described by Hough, Archer & Probert (2018).

## Data analysis

### Microbiome data analysis

The data generated from sequencing of the 16S rRNA gene amplicon libraries were processed using the QIIME pipeline (version 1.8.0; <http://qiime.org/>) (Caporaso *et al.*, 2010b). Sequences from different samples were demultiplexed according to their barcode sequences and chimeric sequences were identified using the UCHIME algorithm (Edgar *et al.*, 2011) and were filtered from the data. Sequences were then clustered open-reference into operational taxonomic units (OTUs) at 97% identity threshold using USEARCH (version 6.1.544) (Edgar, 2010). A representative sequence for each OTU cluster was aligned to the Greengenes core set (version 13.8) (DeSantis *et al.*, 2006) using PyNASt (Caporaso *et al.*, 2010a), filtered to remove gaps and hypervariable regions using the Lane mask before creating an approximately-maximum-likelihood phylogenetic tree using FastTree (Price, Dehal & Arkin, 2010). Taxonomic assignment of OTU representatives was performed using the ribosomal database project classifier (version 2.2) (Wang *et al.*, 2007) informed with the Greengenes reference database at a 50% confidence limit.

Statistical analyses were performed using R software environment (version 3.2.2) (R Core Team, 2014) with the following add-on statistical packages: ‘phyloseq’ (McMurdie & Holmes, 2013), ‘vegan’ (Oksanen *et al.*, 2015), ‘ggplot2’ (Wickham, 2009), ‘nlme’ (Pinheiro *et al.*, 2015), and ‘cluster’ (Maechler *et al.*, 2016). Apart from alpha diversity analysis, the OTU table was further filtered to remove low-abundance OTUs (OTUs present in <5% of samples or represented by <20 reads) and normalised by rarefying to account for unequal sequencing effort between samples (Weiss *et al.*, 2015).

The data were clustered hierarchically based on the average linkage agglomerative clustering method (UPGMA) following calculation of a Bray–Curtis dissimilarity matrix. The trees were visualised using publicly available software (FigTree version 1.4.2). Principal coordinate analysis (PCoA) was performed on a Bray–Curtis dissimilarity matrix created from the OTU table. The amount of variation in the data that could be explained by either the time of sample collection relative to foaling, feed or the individual mares was estimated using permutational multivariate analysis of variance (PERMANOVA)

following calculation of a Bray–Curtis dissimilarity matrix from the data using the ‘vegan::adonis’ function in R.

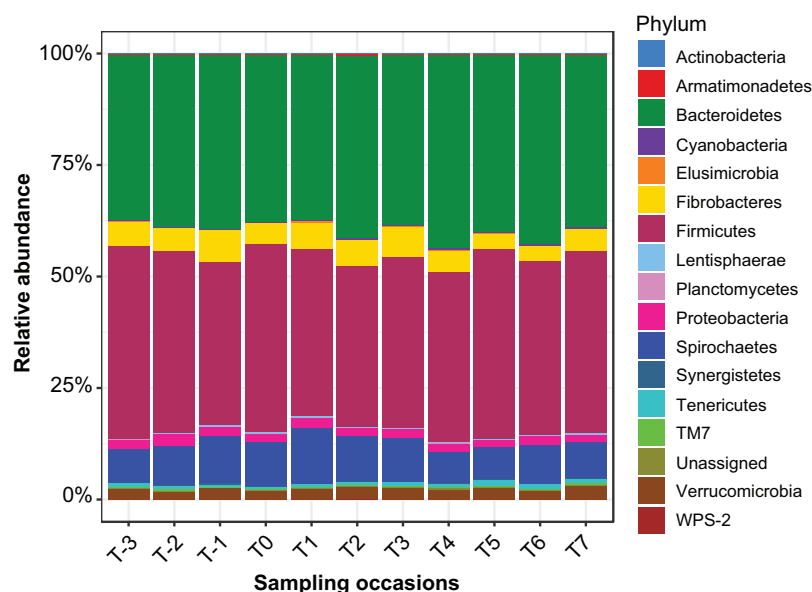
The statistical methods and R script used for investigating changes in faecal microbiota diversity and stability over time in the current study were adapted from publicly available R codes ([DiGiulio et al., 2015a](#)). Alpha diversity analysis involved calculation of Chao1 index ([Chao, 1984](#)) for species richness (number of different species within a community); and Shannon ([Shannon, 1948](#)) diversity index for population diversity (a measure of species richness and similarity of species abundance within a community). The pattern of change of calculated measures over time relative to foaling were evaluated using linear mixed-effects modelling (LME). Random intercept and slope LME models were fitted where mares were included as a random effect and time relative to foaling (in weeks) was included as a fixed effects term in the model.

Beta diversity analysis involved calculation of weighted-UniFrac ([Lozupone et al., 2007](#)) and Bray–Curtis ([Bray & Curtis, 1957](#)) dissimilarity metrics. Distances between consecutive samples of the community (distances between consecutive sampling time points) within each of the mares were calculated and were used as a measure of stability of the community over time relative to foaling. The calculated distances were modelled using random intercept and slope LME models where mares were treated as a random effect variable and where time was the fixed portion of the model.

Samples collected during a 3-week period before and after foaling, were compared for differentially abundant OTUs using negative binomial models. The models were fitted using the DESeq2::DESeq function in R. Prior to differential abundance analysis, the OTU table was further filtered to exclude OTUs present in <25% of the samples. *p*-values were adjusted for multiple testing using the false discovery rate (FDR) method ([Benjamini & Hochberg, 1995](#)). OTUs the adjusted *p*values of which were <0.1 were considered significant. Results from this model were presented in a dot plot.

### **Metabolome data analysis**

Metabolome data were processed using Automated Mass Spectral Deconvolution System (AMDIS-version 2.71, 2012) and VOCs were putatively identified using the National Institute of Standards and Technology mass spectral library (version 2.0, 2011). Data were aligned using the R package Metab ([Aggio, Villas-Boas & Ruggiero, 2011](#)). All samples were analysed in triplicate, an average was taken of the technical replicates and taken forward for data analysis. To allow statistical comparison of relative compound abundance, any missing values present after taking an average were replaced with a half-minimum value of the data matrix. Clustering within the data was investigated using principal component analysis (PCA) and the mean number of VOCs identified were compared between individual mares, weeks of sample collection and the types of feeds using ANOVA followed by Tukey’s HSD test. PERMANOVA was used to estimate the amount of variation in the data that could be explained by individual mares, week of sample collection and the type of feed. LME modelling was used to explore the pattern of change of individual VOCs over time. For each VOC a Random intercept model was fitted where mares were included as a random effect variable and time in weeks was



**Figure 1** Relative abundance of bacterial phyla identified in the mare faecal microbiota. A bar plot of the relative abundance of different bacterial phyla identified in the data.

[Full-size](#) DOI: [10.7717/peerj.6687/fig-1](https://doi.org/10.7717/peerj.6687/fig-1)

included as a fixed effects term in the model. VOCs with  $p$ -values of  $<0.1$  following adjustment for multiple comparisons using the FDR method were considered significant. All analyses were performed using R (version 3.2.2).

## RESULTS

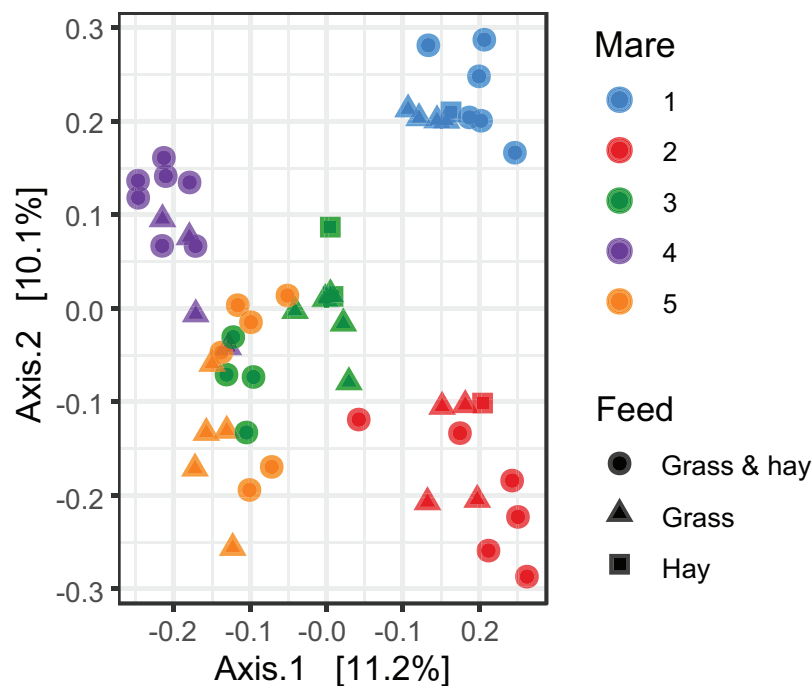
### Mare faecal microbiota

Sequencing of PCR-amplified 16S rRNA genes from 55 samples resulted in 1,648,876 quality non-chimeric sequences. Each sample had at least 17,392 reads, and there were an average of 29,980 reads per sample. The reads were clustered into 17,863 OTUs. Filtration of spurious OTUs reduced this count to 7,843 OTUs (40% of the original count). In this filtered OTU count table, 16 bacterial phyla were identified (Fig. 1), the relative abundance of which at each sampling time point is presented in Table S2.

Cluster analysis revealed that the data were clustered by mares rather than by time relative to foaling. UPGMA trees built from a Bray–Curtis dissimilarity matrix is provided in Fig. S2. A similar pattern of clustering was also confirmed in a PCoA plot (Fig. 2). PERMANOVA analysis revealed that 3% ( $R^2 = 0.03$ ,  $p$ -value = 0.001) of variation in the data could be explained by the variable time, 5% ( $R^2 = 0.05$ ,  $p$ -value = 0.001) by the type of feed and 33% ( $R^2 = 0.33$ ,  $p$ -value = 0.001) by individual mares.

Linear mixed-effects modelling of Chao1 and Shannon index diversity measures revealed general stability of alpha diversity over time. Prediction plots from these models are given in Figs. 3A and 3B. Similar results were also obtained when investigating dissimilarity (beta diversity) between consecutive sampling time points (Figs. 3C and 3D). These results suggest that the mare faecal microbiota was stable from 3 weeks pre-foaling to 7 weeks post-foaling. Only 81 OTUs were found to be significantly differentially





**Figure 2** Principal coordinate analysis of the mare faecal microbiota. Ordination plot of the first two axes from the principal coordinate analysis (PCoA) of microbiome data. Data points are coloured by individual mares and shaped by the types of feed. [Full-size !\[\]\(fcc3264021d438d9732560e78099f674\_img.jpg\) DOI: 10.7717/peerj.6687/fig-2](https://doi.org/10.7717/peerj.6687/fig-2)

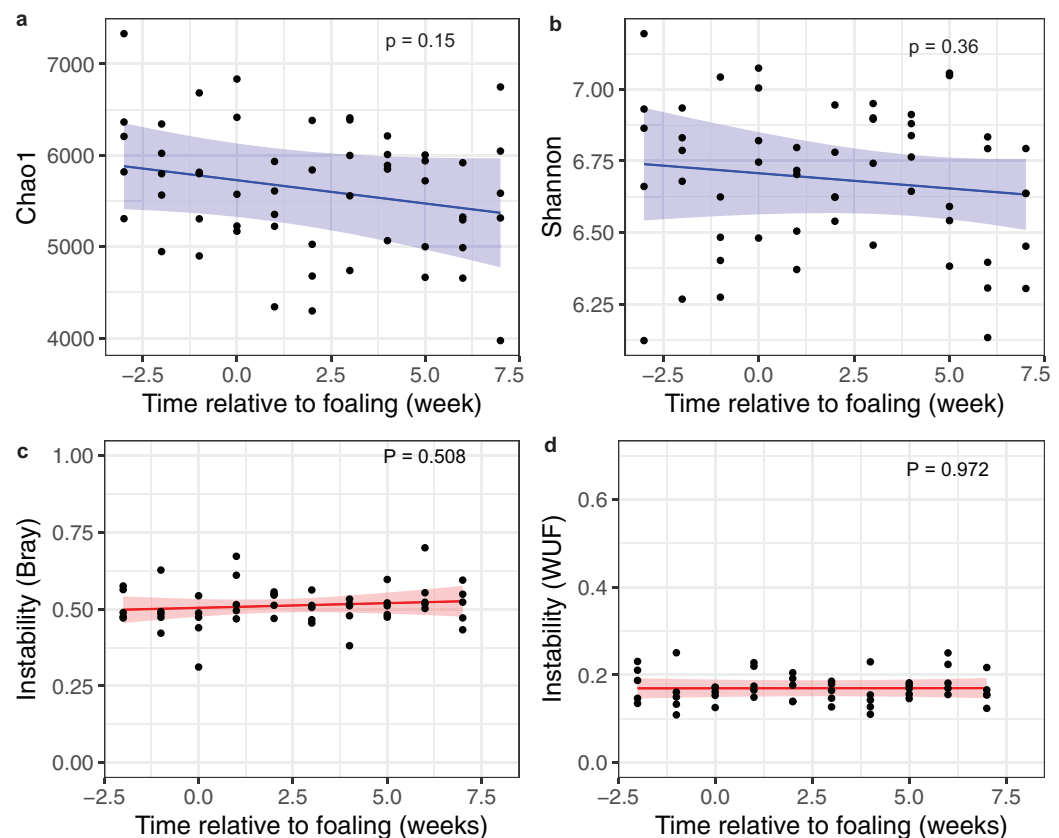
abundant between samples collected during a 3-week period prior to foaling and a 3-week period post foaling (Fig. S3). These results also confirm greater stability of the mare faecal microbiota around the time of foaling. Of these OTUs, 54 were upregulated and 27 were downregulated post-foaling. Downregulated OTUs belonged mainly to the Ruminococcaceae family ( $n = 39$ ) while the upregulated OTUs were represented mainly by members of Spirochaetaceae family ( $n = 6$ ) (Table S3).

### Mare faecal metabolome

A total of 98 VOCs were characterised in the faecal metabolome in all samples. The mean number of VOCs did not differ significantly between individual mares, weeks of sample collection or the types of feed (Figs. S4A–S4C). There was no clear clustering of the data on PCA analysis (Figs. 4 and 5). Similar to the microbiome data, PERMANOVA analysis showed that a small amount of variation could be explained by the time of sampling ( $R^2 = 0.05$ ,  $p = 0.01$ ) and type of feed ( $R^2 = 0.06$ ,  $p = 0.04$ ), while individual mares accounted for 12% of variation in the data ( $R^2 = 0.12$ ,  $p = 0.04$ ). LME modelling identified three VOCs which increased significantly (adjusted  $p$ -value  $< 0.1$ ) over time during the study period (Fig. 6). These were decane, 2-pentylfuran and oct-2-ene.

## DISCUSSION

The current study has characterised the faecal microbiota and the faecal metabolome of a group of pregnant mares during the periparturient period, at times when broodmares are at greatest risk of colic. We found that these microbial communities were structurally



**Figure 3** Changes in alpha and beta diversities over time. Prediction plots of (A) Chao1 and (B) Shannon diversity measures; and (C) a Bray–Curtis and (D) a Weighted Unifrac dissimilarity matrices against time relative to foaling. Blue and red lines are the regression lines from the linear mixed-effects models and cognate shading denotes the 95% confidence limits of the prediction. The models included time as a fixed effects variable and the mares as a random effect variable.

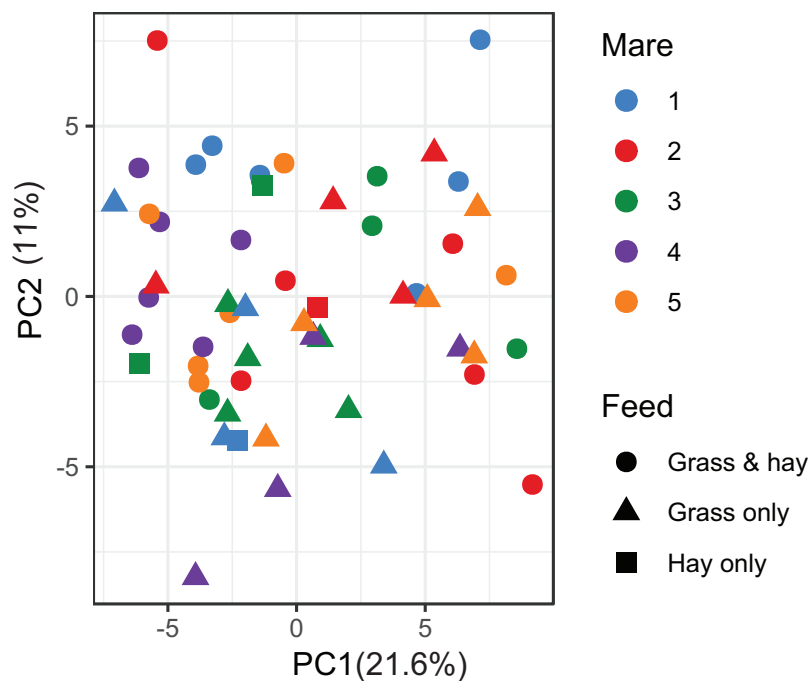
Full-size [DOI: 10.7717/peerj.6687/fig-3](https://doi.org/10.7717/peerj.6687/fig-3)

and functionally stable over the course of the study, and changes that were identified were largely associated with individual mares rather than being related to the time of sample collection. These findings are in agreement with previous studies (*Blackmore et al., 2013; Proudman et al., 2015*) and suggest that if changes in faecal microbiota are associated with postpartum colic, altered risk may be due to inherent differences between individual mares.

The findings of the current study are consistent with a recent human study that investigated the microbiota composition of four different body sites including the distal gut, vagina, saliva, and tooth/gums of pregnant women (*DiGiulio et al., 2015b*). The latter study found that these microbial communities were stable throughout pregnancy and post-delivery. Our findings are also consistent with those reported by *Weese et al. (2015)* who reported a minimal effect of foaling on the mare faecal microbiota. None of the mares included in the current study developed colic either pre- or postpartum and therefore the results reported here cannot be related directly to colic occurrence.

However, our results suggest that it is unlikely that inherent changes in faecal microbiota



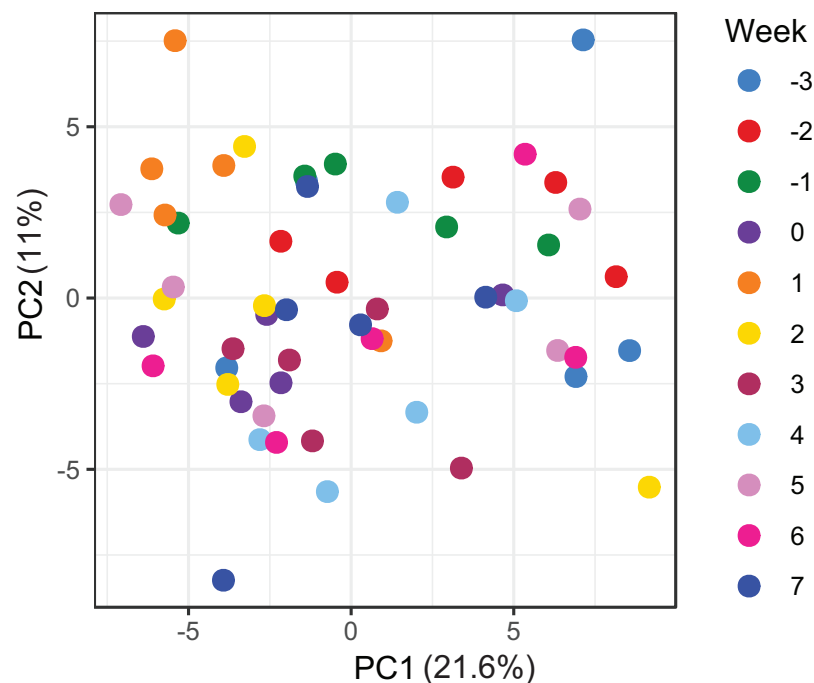


**Figure 4** Principal component analysis of metabolome data. Ordination plot of the first two axes from the principal component analysis (PCA) of metabolome data. The plot shows that there is no clear clustering of the data by either the individual mares or the types feeds available to the mares. Data points are coloured by individual mares and shaped by the types of feeds.

Full-size [DOI: 10.7717/peerj.6687/fig-4](https://doi.org/10.7717/peerj.6687/fig-4)

of mares during the periparturient period are a major contributing factor in the development of periparturient colic. [Weese et al. \(2015\)](#) reported significant increase in relative abundance of the phylum Proteobacteria and reduction in relative abundance of the Firmicutes, Bacteroidetes, and Tenericutes phyla prior to the onset of postpartum colic. This observation requires further investigation to determine if such markers can identify mares at increased risk of colic and any interventions that may reduce this risk.

The level of alpha diversity of the mare faecal microbiota exhibited an overall linear stability with time relative to foaling. Alpha diversity is a measure of within-sample biodiversity and is often used to associate the perturbation of microbial communities with a disease process or changes in the physiological status of the host ([De Weerth et al., 2013](#); [Elli, Colombo & Tagliabue, 2010](#); [Turnbaugh et al., 2009](#)). Changes in beta diversity overtime were also minimal which together suggest that the mare faecal microbiota exhibited minimal change during the period under investigation. The significant increase in mean count of OTUs that belonged to the Ruminococcaceae family and the decrease in members of Spirochaetaceae family post-foaling could be attributed to changes in feed types where mares were predominantly fed on grass post-foaling. These findings concur with a previous study, which reported increase in relative abundance of members of Ruminococcaceae and decrease in relative abundance of members of Spirochaetaceae bacterial families in response to grass feeding ([Salem et al., 2018](#)).

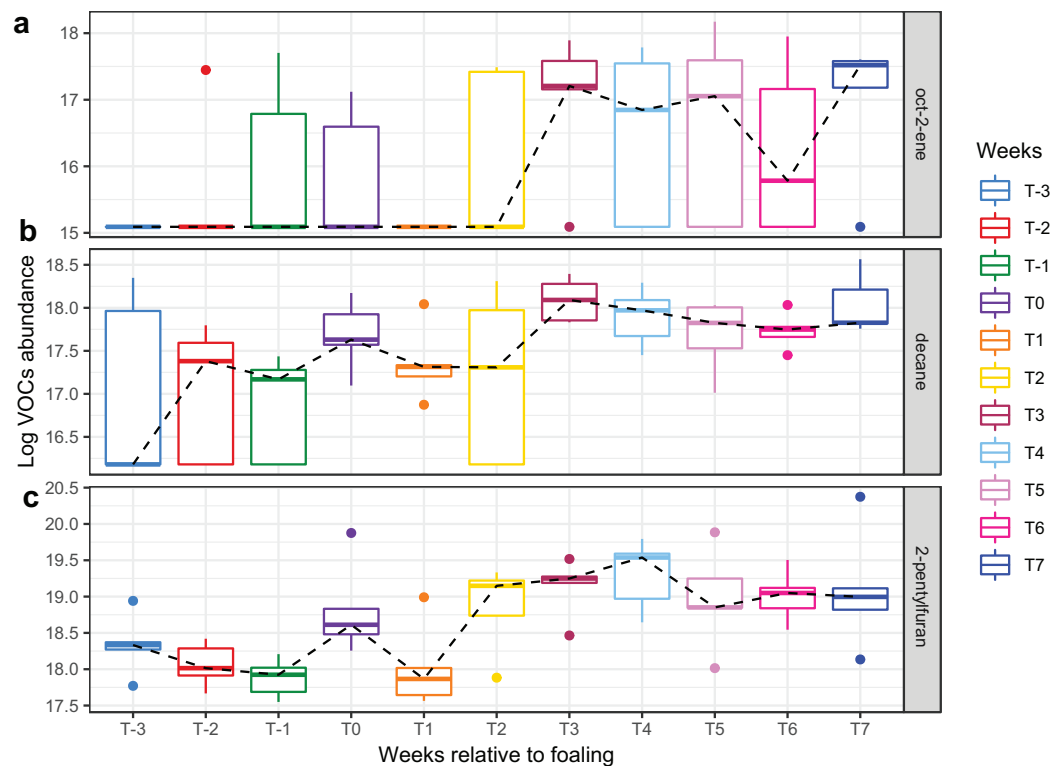


**Figure 5** Principal component analysis of metabolome data. Ordination plot of the first two axes from the principal component analysis (PCA) of metabolome data. The plot shows that there is no clear clustering of the data by sampling time points. Data points are coloured by the time of sampling relative to foaling. [Full-size !\[\]\(1663bb69f307a960345edb0e712f8c02\_img.jpg\) DOI: 10.7717/peerj.6687/fig-5](https://doi.org/10.7717/peerj.6687/fig-5)

Three compounds (decane, 2-pentylfuran, and oct-2-ene) altered significantly with time in the current study, while the remainder of the faecal VOC metabolome of the mares remained stable.

Decane, 2-pentylfuran, and oct-2-ene have been detected from the faecal headspace of ruminants and they may originate directly from food or as products of microbial fermentation (Cai *et al.*, 2006; Fischer *et al.*, 2015; Garner *et al.*, 2007; Schulz & Dickschat, 2007). Therefore, the significant changes in these VOCs in the current study may reflect a change in feed type or mare feed intake over time. In general, the faecal VOC metabolome appeared to be more uniform among individual mares than the faecal microbiota, indicating a common functionality between bacteria. Overall, the microbiome and metabolome mirrored each other, in that they both remained stable, indicating that VOCs may be a cost-effective alternative to monitoring the faecal microbiome.

In the current study, the faecal microbiota was used as a proxy for the hindgut microbial populations. Studies that compared the microbiota composition of different regions of the horse gastrointestinal tract have found that faecal microbiota partially represent the microbial populations of the large intestine, particularly those of the distal colon (Costa *et al.*, 2015; Dougal *et al.*, 2012). Given that it is not possible to directly sample the colonic microbiota sequentially, faecal analysis is the best possible proxy measure of potential changes in the colonic microbial community. The effect of a new diet on the equine gut microbiota might not be evident until 4–6 days following dietary change (Fernandes *et al.*, 2014; Van Den Berg *et al.*, 2013). Ideally, mares maintained on one diet



**Figure 6** Boxplots of three VOCs that significantly increased over time during the study period. Boxplots of three VOCs ((A) oct-2-ene, (B) decane, (C) 2-pentylfuran) that significantly increased over time during the study period. Changes of VOC abundance over time was modelled using linear mixed-effects modelling with the mare included as a random effect variable and the time (in weeks) as a fixed effects variable in the model. [Full-size !\[\]\(ba1b80118482ccef74a5d718ca4d7242\_img.jpg\) DOI: 10.7717/peerj.6687/fig-6](https://doi.org/10.7717/peerj.6687/fig-6)

could have been compared to groups of mares on differing diets. This was not possible in the current study nor does it reflect how broodmares are generally managed on many stud farms. Transition between these diets was abrupt, and it was not known exactly when the faecal microbiota changed in relation to this.

## CONCLUSIONS

This is the first study to perform a detailed investigation of faecal microbiota and faecal metabolome of a group of mares during the periparturient period when they are at increased risk of colic. The study demonstrated that the majority of changes identified in the faecal microbiota and VOCs were mare-specific, and did not appear to be related to inherent physiological changes associated with foaling. The change in three VOCs post-parturition is interesting and may reflect a subtle change in the functional microbiome. Further studies are warranted to identify mares with altered metabolome in the periparturient period to determine whether there is a link to the risk of colic. VOCs may provide a cost-effective means of monitoring such mares with the ultimate aim of developing stable-side tests to identify and monitor mares at increased risk of colic.

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## ADDITIONAL INFORMATION AND DECLARATIONS

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### Grant Disclosure

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The Egyptian Ministry of Higher Education.

### Competing Interests

The authors declare that they have no competing interests. Sarah J. Stoneham was employed by the Cheshire Equine Clinic during the period of sample collection and she is currently employed as an equine nutritionist by TopSpec Equine Ltd. North Yorkshire, UK.

### Author Contributions

- Shebl E. Salem conceived and designed the experiments, performed the experiments, analysed the data, prepared figures and/or tables, authored or reviewed drafts of the paper, approved the final draft.
- Rachael Hough conceived and designed the experiments, performed the experiments, analysed the data, authored or reviewed drafts of the paper, approved the final draft.
- Chris Probert conceived and designed the experiments, contributed reagents/materials/analysis tools, authored or reviewed drafts of the paper, approved the final draft.
- Thomas W. Maddox conceived and designed the experiments, approved the final draft.
- Philipp Antczak analysed the data, contributed reagents/materials/analysis tools, approved the final draft.
- Julian M. Ketley conceived and designed the experiments, contributed reagents/materials/analysis tools, approved the final draft.
- Nicola J. Williams conceived and designed the experiments, contributed reagents/materials/analysis tools, approved the final draft.
- Sarah J. Stoneham conceived and designed the experiments, approved the final draft.
- Debra C. Archer conceived and designed the experiments, authored or reviewed drafts of the paper, approved the final draft.

## Animal Ethics

The following information was supplied relating to ethical approvals (i.e., approving body and any reference numbers):

The University of Liverpool Veterinary Research Ethics Committee approved the study (VREC207).

## Data Availability

The following information was supplied regarding data availability:

The raw sequence data, Qiime script and the R codes used for data analyses are available at figshare: Salem, Shebl (2019): A longitudinal study of the faecal microbiome and metabolome of periparturient mares. figshare. Fileset. DOI [10.6084/m9.figshare.7267343.v1](https://doi.org/10.6084/m9.figshare.7267343.v1).

Sequence data are available at the NIH Sequence Read Archive, accession number [PRJNA523602](https://www.ncbi.nlm.nih.gov/sra/PRJNA523602).

## Supplemental Information

Supplemental information for this article can be found online at <http://dx.doi.org/10.7717/peerj.6687#supplemental-information>.

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