Escherichia coli O157:H7 Super-Shedder and Non-Shedder Feedlot Steers Harbour Distinct Fecal Bacterial Communities

Yong Xu1, Eric Dugat-Bony2, Rahat Zaheer3, Lorna Selinger3, Ruth Barbieri3, Krysty Munns3, Tim A. McAllister3, L. Brent Selinger1

1 Department of Biological Sciences, University of Lethbridge, Lethbridge, Alberta, Canada, 2 AgroParisTech National Institute for Agricultural Research, Thiverval, Grignon, France, 3 Agriculture and Agri-Food Canada, Lethbridge Research Centre, Lethbridge, Alberta, Canada

Abstract

Escherichia coli O157:H7 is a major foodborne human pathogen causing disease worldwide. Cattle are a major reservoir for this pathogen and those that shed E. coli O157:H7 at >10⁴ CFU/g feces have been termed “super-shedders”. A rich microbial community inhabits the mammalian intestinal tract, but it is not known if the structure of this community differs between super-shedder cattle and their non-shedding pen mates. We hypothesized that the super-shedder state is a result of an intestinal dysbiosis of the microbial community and that a “normal” microbiota prevents E. coli O157:H7 from reaching super-shedding levels. To address this question, we applied 454 pyrosequencing of bacterial 16S rRNA genes to characterize fecal bacterial communities from 11 super-shedders and 11 contemporary pen mates negative for E. coli O157:H7. The dataset was analyzed by using five independent clustering methods to minimize potential biases and to increase confidence in the results. Our analyses collectively indicated significant variations in microbiome composition between super-shedding and non-shedding cattle. Super-shedders exhibited higher bacterial richness and diversity than non-shedders. Furthermore, seventy-two operational taxonomic units, mostly belonging to Firmicutes and Bacteroidetes phyla, were identified showing differential abundance between these two groups of cattle. The operational taxonomic unit affiliation provides new insight into bacterial populations that are present in feces arising from super-shedders of E. coli O157:H7.

Citation: Xu Y, Dugat-Bony E, Zaheer R, Selinger L, Barbieri R, et al. (2014) Escherichia coli O157:H7 Super-Shedder and Non-Shedder Feedlot Steers Harbour Distinct Fecal Bacterial Communities. PLoS ONE 9(5): e98115. doi:10.1371/journal.pone.0098115

Editor: Athol Victor Klieve, University of Queensland, Australia

Received February 10, 2014; Accepted April 28, 2014; Published May 23, 2014

Copyright: © 2014 Xu et al. This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Funding: The project was funded by Genome Alberta (genomealberta.ca). The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing Interests: The authors have declared that no competing interests exist.
* E-mail: selibl@uleth.ca
† These authors contributed equally to this work.

Introduction

Escherichia coli O157:H7 is a serotype of enterohemorrhagic E. coli (EHEC) that produces shiga-like toxins and is a major public health concern worldwide. Conditions caused by E. coli O157:H7 range from mild diarrhea to hemorrhagic colitis and life-threatening hemolytic uremic syndrome [1]. This pathogen typically affects children, the elderly and immunocompromised patients [2]. Cattle are a principal reservoir of E. coli O157:H7, which preferentially colonizes the lower gastrointestinal tract (GIT) of cattle and in particular the rectal-anal junction [3,4]. Cattle can shed E. coli O157:H7 heterogeneously, ranging from some individuals that do not shed E. coli O157:H7 (non-shedders) to those that shed very high concentrations of the pathogen. Cattle shedding E. coli O157:H7 >10⁴ CFU/g of feces have been designated as “super-shedders” [5]. Super-shedders are hypothesized to be responsible for majority of transmission of this bacterium among individuals and its dissemination into the environment. Consequently, they have been proposed as prime targets for mitigation strategies aimed at reducing the incidence and spread of E. coli O157:H7 [6].

Currently, very little is known about the factors that contribute to the shedding of E. coli O157:H7. Microbial-specific, host-specific and environmental factors likely all play a role in the development of the super-shedding phenomenon. Based on many recent studies on the impact of intestinal microbiota on host well-being, it could be hypothesized that the incidence and shedding of E. coli O157:H7 is likely affected not only by characteristics of the strain but also by the nature of the microbiome (i.e., bacteria, fungi and protozoa) within the host’s GIT. The GIT microbial community is critical to host health and well-being and it is projected that even minor changes in these populations may cause dramatic shifts that affect livestock productivity [7]. The normal microbiota not only produces necessary nutrients (i.e., vitamins and short chain volatile fatty acids) [8,9] but also has been implicated in the development of a healthy immune system and the exclusion of enteric pathogens [10]. Recent studies using fecal samples from cattle have revealed a high degree of animal-to-animal variation in the GIT bacterial community [11,12].

In this study, we hypothesized that the super-shedder condition is a result of intestinal dysbiosis and that a healthy normal microbiota prevents E. coli O157:H7 from colonizing the intestinal...
tract in a manner that leads to the super-shedding state. To test this, we employed Bacterial Tag-Encoded FLX Amplicon Pyrosequencing Analysis (bTEFAP) [13] to generate over 200,000 partial 16S rRNA gene sequences from fecal samples collected from 11 super-shedders (SS cattle) and 11 contemporary pen mates which were negative for E. coli O157:H7 (NS cattle). Sequence data was processed to 1) depict the bacterial community structure for fecal samples from SS cattle, 2) compare the bacterial community profiles for SS and NS cattle and 3) identify the operational taxonomical units (OTUs) differentially present in each group.

Materials and Methods

Sample collection and E. coli O157:H7 enumeration

All cattle used in this experiment were handled in accordance with the Canadian Council of Animal Care [14] and the protocol (#1120) was reviewed and approved by the nationally accredited Lethbridge Research Centre Animal Care Committee. Crossbred yearling feedlot steers (n = 400), from a single commercial feedlot in southern Alberta were sampled in July of 2011. All cattle were fed a barley-grain based finishing diet. Fecal samples (~50 g) were collected by rectal palpation, immediately placed on ice, transported to the laboratory and analyzed within 4 h after collection. E. coli O157:H7 was enumerated from feces by serially diluting 1 g into 9 mL of phosphate buffered saline and plating E. coli O157:H7 (NS cattle). Sequence data was processed to 1) depict the bacterial community structure for fecal samples from SS cattle, 2) compare the bacterial community profiles for SS and NS cattle and 3) identify the operational taxonomical units (OTUs) differentially present in each group.

Bacterial Tag-Encoded FLX Amplicon Pyrosequencing Analysis

Community DNA was extracted from 0.25 g of feces from SS cattle (n = 11) and NS cattle (n = 11) using the QIAamp DNA Stool kit (Qiagen, Toronto, ON, Canada) and analyzed at the Research and Testing Laboratory (Lubbock, Texas, USA) by using bacterial tag-encoded FLX amplicon pyrosequencing (bTEFAP) [13]. Primers that span the hypervariable regions V1 and V3 of the 16S rRNA gene (20F 5’GAGT TTGA-TCNTG GCTCAG and 519R 5’ GTNTACNGCGGCKG-CTG), were used in sequencing reactions. Sequence coverage was approximately 10,000 sequences per sample.

Sequence data processing

Raw sequence read data were quality-filtered using Mothur v.1.23 [16] to remove sequences less than 200 nucleotides (nt), sequences containing homopolymers >8 nt, mismatches in the barcode or primer, >1 ambiguous nt, or regions in which an average quality score below 30 was obtained using a moving window of 50 nt. The remaining sequences were aligned to the SILVA-based bacterial reference alignment [17] using the Needleman-Wunsch algorithm [18]. Potential chimeric sequences were removed using UCHIME in reference mode [19]. Finally, sequencing noise was reduced by a pre-clustering step implemented in Mothur v.1.23 [16]. Sequence data have been submitted to the short-read archive (Bio Project SRP040475 – accession numbers for individual animal samples sequence data are SRX98462, SRX98463, SRX98466, and SRX98471 – SRX98489).

Microbial community analysis

Previous studies have shown that alpha and beta diversity analyses are very sensitive to the clustering methods used for binning sequences into OTUs [20,21]. Therefore, five cluster-based algorithms were used in order to reduce bias: Mothur [16], Espirit-tree [22], Otupipe (http://drive5.com/otupipe/), CROP [23] and UPARSE [24]. These algorithms represent three distinct clustering methods: hierarchical clustering (Mothur, Espirit-tree), greedy heuristic clustering (Otupipe, UPARSE) and Bayesian clustering (CROP). For Mothur an average neighbor algorithm cutoff was used. Default parameters were applied for CROP [23] and Espirit-tree [22]. Two MinSizes were set for Otupipe: MinSize = 2 and MinSize = 4. In order to use the UPARSE pipeline, the qual and fasta files for each sample were combined into fastq files [24]. Sequences were filtered to a fixed length of 180 nt and those with an expected error probability greater than 0.5 were discarded. Chimeras were checked against the “Gold” database based on UCHIME [19], and reads were clustered at 97% sequence identity into operational taxonomic units (OTUs).

For all clustering methods a dissimilarity level of 0.03 was used to bin OTUs, a resolution that corresponds to the species level. To assess similarity among the different methods, two reference OTUs datasets were built by mapping quality-filtered, processed sequences against the Greengenes reference OTUs database [25] downloaded from http://greengenes.lbl.gov/Download/Sequence_Data/Fasta_data_files/Caporaso_Reference_OTUs/gg.otus_4feb2011.tgz using USEARCH [26] with a global identity threshold of 97% and 90%, respectively. Taxonomic assignments of OTUs were determined using the Ribosomal Database Project Bayesian classifier algorithm (version 2.2) [27], with an 80% support cutoff. The most recent taxonomy annotation [28] was used to retrain the ribosomal database project (RDP) classifier.

Statistical analysis

The pyrotag libraries for each sample were normalized based on subsampling to the lowest number of reads (16; Table 1). The different OTU matrices obtained by the above methods were imported into the R version 2.14 statistical computing environment (http://www.r-project.org/). Subsequent analyses including Good’s coverage, Richness and Shannon Diversity Index and
visualization were performed in R using functions from the picante version 1.3 [29], vegan version 2.03 [http://r-forge.r-project.org/projects/vegan/] , phyloseq version 1.14 [30] and ggplot2 version 0.8.9 [31]. Bacterial community dissimilarity was quantified using Bray-Curtis distance. The Bray-Curtis distance matrix was used to test if the bacterial communities of non-shedders and super-shedders were statistically different by running the ANOSIM function in the vegan package with 1000 permutations. A non-parametric Wilcoxon rank sum test was implemented by R script with *p* < 0.05 and False Discovery Rate (FDR) < 25% to identify OTUs with significant differences between non-shedders and super-shedders. The compositional similarity of all samples was visualized using a nonmetric multidimensional scaling (NMDS) ordination.

**Results**

**E. coli O157:H7 shedding levels by feedlot cattle**

From sampling 400 cattle at the commercial feedlot, a total of 46 (11.5%) were identified as shedding **E. coli O157:H7** of which 11 (2.8%) of these cattle were super-shedders (>10^8 **E. coli O157:H7** cfu/g feces). **E. coli O157:H7** counts for the super-shedders ranged from 1.5 × 10^7 to 6.5 × 10^7 cfu/g feces. Eleven non-shedding pen mates were also selected as the control group.

**General sequencing information and taxonomic affiliation**

Next generation pyrosequencing was used to sequence 16S rRNA genes in metagenomic DNA extracted from fecal samples collected from the 11 SS cattle and 11 NS cattle. In total 236,139 pyrotags were obtained from the 22 fecal metagenomic DNA samples. After removal of low quality reads and chimeras, 112,926 processed sequences remained, with an average of 5,126 reads per sample (ranging from 2,942–7,622) and a mean length of 278 nt (Figure S1). Based on an 80% RDP confidence level, 18 phyla, 33 classes, 57 orders, 77 families and 97 genera were inferred. As expected, the proportion of unclassified sequences increased as the taxonomic grouping decreased. At the phylum level, only 4.4% (4,914) of the sequences remained unclassified, while at the genus level 76.7% (86,591) of the sequences could not be classified. For all samples, the bacterial communities were dominated by **Firmicutes**, accounting for 53.9% of the dataset (60,845/112,926 sequences), followed by **Bacteroidetes** (35.6%; 40,185 sequences), **Spirochaetes** (3.7%; 2,521 sequences) and **Proteobacteria** (1.7%; 1,892 sequences) (Figure 1). The remaining 14 phyla accounted for less than 2.3% of the dataset. At the genus level, **Prevotella (Bacteroidetes phylum)** was identified most frequently in fecal samples, accounting for up to 10.5% of the entire dataset. All of the taxa within each of the five taxonomic levels are listed in Table S1. The classifier-based analysis revealed that the composition of bacterial communities differed among steers. Only 5/18 phyla (**Bacteroidetes, Firmicutes, Proteobacteria, Spirochaetes, Tenericutes**), 8/33 classes (**Bacilli, Bacteroidia, Betaproteobacteria, Clostridia, Erysipelotrichi, Gammaproteobacteria, Mollicutes, Spirochaetes**), 5/57 orders (**Bacteroidales, Bacillales, Clostridiales, Erysipelotrichales, Spirochaetales**), 12/77 families (**Alcaligenaceae, Bacteroidaceae, Clostridiaceae, Clostridiales Family XIII miteae sedis, Erysipelotrichaceae, Lachnospiraceae, Peptococcaceae, Prevotellaceae, Rikenellaceae, Ruminococcaceae, Spirochaetaceae, Veillonellaceae**) and 9/97 genera (**Alistipes, Bacteroides, Clostridium, Eubacterium, Oscillospira, Prevotella, Ruminococcus, Sutterella, Treponema**) were shared by all 22 samples.

**Super-shedder cattle exhibit more diverse fecal bacterial communities than non-shedder cattle**

Previous studies have shown that alpha and beta diversities are very sensitive to the clustering methods used for binning sequences into OTUs [20,21]. Therefore, in this study, five cluster-based algorithms were used to reduce bias. In addition we built two OTU matrices, RefMap97 and RefMap90 by mapping the pyrotags to the Greengenes reference OTUs with global sequence similarities of 97% and 90%, respectively (Table 1). The reference OTU matrices, RefMap97 and RefMap90, are based on the full-length 16S rRNA genes with strict quality control.

Only 38.6% (97% similarity) and 84.3% (90% similarity) of the OTUs could be mapped onto the Greengenes reference OTUs. Five clustering methods that were used to bin pyrotags into OTUs are summarized in Table 1. Three methods including Mothur, Esprit-tree and CROP were able to use all of the processed sequences. In comparison, UPARSE used 34.5% of the processed sequences and Otipipe used 37.6% and 83.1% of the sequences at MinSize parameters of 2 and 4, respectively. The Otipipe MinSize parameter sets the minimum size of a cluster in the precluster process resulting in some sequences being discarded as noise. The total number of OTUs utilized among clustering methods differed by >10 fold.

The alpha diversity results for each clustering method exhibited similar trends; consequently, only those obtained with the Esprit-tree method are presented in Figure 2. While the rarefaction curves did not plateau (Figure 2A), Good’s coverage (which estimates the probability that the next read will belong to

**Table 1. Comparison of OTU-binning methods**

| Method        | No. of OTUs | Pyrotags binned into OTUs | No. of pyrotags for normalization | No. of OTUs after normalization | Correlation value to RefMap97 |
|---------------|-------------|----------------------------|-----------------------------------|--------------------------------|-----------------------------|
| Mothur        | 8,664       | 100%                       | 2,942                             | 6,823                          | 0.77                        |
| Esprit-tree   | 5,701       | 100%                       | 2,942                             | 4,733                          | 0.79                        |
| CROP          | 1,427       | 100%                       | 2,942                             | 1,246                          | 0.83                        |
| UPARSE        | 1,112       | 34.5%                      | 859                               | 941                            | 0.92                        |
| Otipipe2*     | 3,801       | 83.1%                      | 2,502                             | 3,540                          | 0.91                        |
| Otipipe4*     | 1,111       | 37.9%                      | 1,091                             | 926                            | 0.96                        |
| RefMap90*     | 1,780       | 84.7%                      | 2,419                             | 1,545                          | 0.94                        |
| RefMap97*     | 862         | 37.8%                      | 968                               | 728                            | 1.00                        |

*Otipipe2: MINSIZE = 2 (i.e., the minimum depth is 2); Otipipe4: MINSIZE = 4. RefMap90: the sequences were mapped into the reference database with an identity threshold of 90%. The identity threshold was 97% for RefMap97. No. of pyrotags with the smallest library for the normalization.

doi:10.1371/journal.pone.0098115.t001
existing OTU) averaged 86.9\%±3.3\% (mean ± SD) (Figure 2B), indicating that the majority of bacterial diversity was captured with this sequencing effort. The richness (Figure 2C) and Shannon diversity index (Figure 2D) were greater \((p<0.05, \text{Wilcoxon test})\) for SS cattle than for NS cattle, regardless of the clustering method used to process the data. Together, these results highlight the fact that fecal bacterial communities of super-shedding cattle are more diverse than those of non-shedders.

Super-shedders and non-shedders have distinct bacterial community structures

The SS and NS groups formed separate clusters on NMDS ordinate plots (Figure 3) for the majority of OTU matrices generated in this study. Analysis of similarity (ANOSIM) further confirmed these differences for six of the eight OTU matrices: CROP, UPARSE, Otupipe2, RefMap90 and RefMap 97; \((p\leq 0.05)\) and Otupipe4 \((p \leq 0.01)\). These results provide strong support that the fecal bacterial communities of SS and NS groups had greater inter-group than intra-group differences. Furthermore, the Bray-Curtis distance was significantly larger within NS group than SS group: 1.2 and 0.9 respectively (one tail t-test \(p<0.01\)). Collectively, these results support the contention that SS and NS clearly harbored different fecal bacterial communities.

The Mantel test was used to examine the similarity of distance matrices generated by the different OTU methods; all matrices were highly correlated to the RefMap97 \((p<0.001)\), but to different degrees (Table 1): Otupipe2 was the most similar, followed by RefMap90, UPARSE, Otupipe2, CROP, Mothur and Esprit-tree. Therefore, the Otupipe2 matrix was considered as the
most relevant OTU matrix to differentiate taxa between super shedder and non-shedders owing to its high correlation with the RefMap97 matrix and its ability to cover 83% of processed sequences (Table 1).

Identification of OTUs exhibiting differential abundance in super-shedders and non-shedders

Comparison of DNA extracted from fecal samples arising from SS and NS cattle revealed global significant differences in bacterial community structure. The differential abundance of OTUs was determined with the Otupipe2 OTU matrix and 72 OTUs showed differential abundance ($p<0.05$) between the SS and NS (Figure 4). Seventeen OTUs (23.6%) were enriched in NS, whereas 55 (76.4%) were more abundant in SS. These OTUs included Firmicutes (n = 53), Bacteroidetes (n = 16), Tenericutes (n = 2) and Proteobacteria (n = 1) (Table S2).

Of the differentially abundant Firmicutes OTUs, all were assigned to the Clostridiales order, with four known families represented: Ruminococcaceae (36 OTUs), Lachnospiraceae (10 OTUs), Clostridiaceae (3 OTUs) and Peptococcaceae (1 OTU). Three OTUs...
(OTU_662, OTU_2712, and OTU_2803) remained unclassified at the family level. Eleven OTUs were significantly more abundant in NS compared to 42 OTUs in SS. Among OTUs that were classified at the genus level, *Ruminococcus* OTUs were associated with both NS and SS; OTU_2565 was more often found in NS with the opposite response found for OTU_2018. Of the three *Clostridium* OTUs, two (OTU_2484 and OTU_2680) were more frequently found in SS while the third (OTU_2831) was more abundant in NS. The single *Blautia* (OTU_2750) and *Oscillospira* (OTU_1054) OTUs were more often present in SS.

For the *Bacteroidetes* phylum, all were classified in the *Bacteroidales* order, with the remaining two OTUs not being classified beyond the phylum level. Thirteen OTUs were significantly more abundant in SS compared to 3 OTUs in NS. Eight OTUs classified at the genus level were preferentially abundant in SS: with seven OTUs assigned to *Prevotella* (OTU_333, OTU_386, OTU_700, OTU_748, OTU_1560, OTU_2540 and OTU_2651) and one to *Alistipes* (OTU_1765).

Differentially abundant OTUs identified as *Tenericutes* and *Proteobacteria*, were represented by two OTUs (OTU_446 and

---

**Figure 3.** Non-metric multidimensional scaling plots of the distances for 22 fecal samples based on the different OTU matrices.

Mothur, Esprit-Tree, CROP, UPARSE and Otupipe were used to bin 16S pyrotags into OTUs. Otupipe2 and Otupipe4 matrices were generated with MinSize parameter set to 2 or 4, respectively. In addition, 16S pyrotags were mapped into against a full-length 16S reference dataset to generate OTU matrices at 97% (RefMap97) or 90% (RefMap90) identity. The plots contain data from super-shedders (SS) and non-shedders (NS). ANOISM was used to test for similarities between SS and NS steers, generating correlation (R) and significance (p) values.

doi:10.1371/journal.pone.0098115.g003
OTU_1279) and one OTU (OTU_2365), respectively. These OTUs were more abundant in NS steers.

**Discussion**

In this study, 2.8% of the steers were super-shedding at the time of sampling. This is consistent with findings by Stephens et al., [32] whereby <10% of cattle in a herd were identified as super-shedders. However, in other studies, prevalence has been documented at over 20% [33]. The duration that an individual continues to express the super-shedding trait is unknown and intensive sampling of individual cattle has shown that the levels of *E. coli* O157:H7 in feces are highly variable [34]. At this point it is not known as to what extent this variability may arise from shifts in the resident microbial communities within the digestive tract.

Numerous studies have demonstrated a correlation between microbiota and host health including obesity [35,36], brain development [37] and intestinal disease. In this study, we demonstrated significant differences in fecal bacterial communities between SS and NS cattle. Global differences in the microbiota in feces from SS and NS cattle were clearly visualized by their separation by NMDS based on five different OTU binning methods and eight OTU matrices (Figure 3). Although different OTU binning methods led to very different results, the general inference remained identical, increasing the robustness of our conclusions. We observed an increased richness (i.e., number of observed OTUs) of fecal bacterial communities in steers shedding high levels of *E. coli* O157:H7 and identified 72 OTUs showing significant differential abundance between SS and NS (Figure 4). Contingent upon their ability to be isolated, these bacterial species could potentially be good candidates for genomic analyses and co-culture studies with *E. coli* O157:H7 to explore possible mechanisms of bacterial interaction.

The statistical power needed to detect differential features for OTUs relies heavily on the number of sequences representing each OTU. Thus it is not surprising that most differentially abundant OTUs (i.e., 69 out of 72) are members of the two dominant phyla (*Firmicutes* and *Bacteroidetes*), which combined accounted for 89.5% of the dataset. However, the less abundant taxa may also play a role in the development of the SS state. Two
OTUs from the phylum Tenericutes, which accounted for 0.2% of the total data, were also identified as differentially abundant between SS and NS steers. Although with the current sequencing depth, we can see a general trend of alpha and beta diversity between SS and NS, it is still difficult to recover all of the differentially abundant features. Despite the high sequencing depth (average of 10,000 sequences per sample) and Good’s coverage (average of 86.9%), more than half of the OTUs were represented by less than 5 sequences.

Microbial populations in cattle feces are highly diverse from individual to individual. Our dataset was estimated to cover more than 85% of the bacterial diversity present in feces (Figure 2). The bacterial community in all steers was dominated by Firmicutes, accounting for 53.9%, followed by Bacteroidetes at 35.6% of total sequences, a pattern observed by others [11,12,38]. With this depth, however, only a small proportion of OTUs (<1%, or 10–50 OTUs) were shared by all 22 samples at 97% similarity level (Figure 5; Table S3). Shanks et al. [12] obtained similar results with fecal samples from 30 cattle using 16S rRNA gene pyrosequencing, in which only 9/9,201 OTUs (0.01%) were shared among all samples at a similarity level of 97%. In another study using full length Sanger sequencing of 16S rRNA gene, Durso et al., [11] identified 24 common OTUs out of 1, 906 OTUs (1.3%) at a 97% similarity level in six cattle fecal samples. Good’s coverage and the rarefaction curve showed that the majority of OTUs in this study were captured in the sample analysis (Figure 2), indicating that the small number of shared OTUs is not due to a lack of sequencing depth. Therefore, there is an inherent high degree of variation in the bacterial community in cattle feces, but they clearly differ from those in the feces of humans, chicken, swine, geese or ducks (Figure S2).

In E. coli O157:H7, two systems including the locus of enterocyte effacement (LEE), responsible for establishing attachment and effacement lesions at the recto-anal junction, and the glutamate decarboxylase (gad) acid-resistance system, have been shown to play a critical role in intestinal colonization in cattle [39,40]. Enterohaemorrhagic E. coli (EHEC) use several quorum sensing systems for intercellular signaling including the LuxR homolog SdiA that senses acyl-homoserine lactones (AHLs) which are present in the rumen but not in the lower GIT of cattle as the pH there does not favor AHL [41–44]. Although EHEC do not produce AHL, it is required for colonization of the GIT as it binds and stabilizes SdiA protein. Chemical signaling through SdiA-AHL promotes the survival of EHEC within the ruminant gut as it activates the expression of gad which aides in survival within the acidic rumen and conserves energy by repressing expression of LEE [45,46]. Certain members of the Bacteroidetes have been reported to produce AHLs [47] and it would be interesting to investigate if the bacteria associated with OTUs depicting differential abundance are AHL producers.

Recent studies revealed the profound impact of diet on microbial community composition in cattle. For example, the
abundance of the family Ruminococcaceae of the phylum Firmicutes and the genus Prevotella of the phylum Bacteroidetes were correlated with dietary change [12,48,49]. In our study, all steers were fed the same diet suggesting that community structure differences between SS and NS are linked to factors other than just diet composition. A large number of studies [11,12,39,49], including this one, have observed a high level of inter-animal variability in microbial intestinal communities. This phenomenon coincides with the possibility that conditions within intestinal microbial communities that are conducive to the shedding of high numbers of E. coli O157:H7 only occur in a limited number of cattle within a herd. The NMDS plots (Figure 3) show that bacterial communities from SS clustered together more so than those from NS, although some outliers were evident. The more similar microbial community structures within SS cattle suggest that this specific microbial composition may have allowed E. coli O157:H7 to proliferate and flourish. Perhaps the changes in microbial community composition in SS cattle result in differential degradation of organic matter, leading to a nutritional environment that is more favorable for the proliferation of E. coli O157:H7. Indeed, this enteric pathogen has evolved a number of unique nutritional pathways that may enable it to occupy a niche different from commensal generic E. coli. For example, E. coli O157:H7 strains are more likely to use dulcitol, sucrose and L-galactonic acid-δ-lactone as carbon sources than other commensal E. coli [50]. In addition, E. coli O157:H7 can utilize free ethanolamine in the bovine small intestine as a nitrogen source, a gene cluster that is generally absent in the genomes of most of other bacterial species within gut microbiota [51]. The presence or absence of ethanolamine utilizing bacteria could also be a contributing factor towards the existence of distinct microbial communities between SS and NS cattle.

Another hypothesis to explain the differences between SS and NS cattle takes into account competitive inhibition mechanisms among E. coli strains [52,53]. Some E. coli strains are able to inhibit the growth of others, resulting in shifts in E. coli populations that could influence the density of E. coli O157:H7 within the GIT. In this study, E. coli sequences represented only 0.015% of the dataset (17 pyrotags) and 16S pyrosequencing analysis was unable to discriminate between strains of E. coli. Further comparative studies focusing on E. coli populations could be useful to evaluate the importance of commensal strains in SS and NS cattle. Interestingly, the genome of E. coli O157:H7 is 25% larger and contains ~1400 more genes than generic E. coli K-12 [54]. Among these genes, many are related to virulence but the function of >60% is unknown, but they may confer specific traits that increase the competitiveness of E. coli O157:H7 within the GIT of cattle. The recent sequencing of 26 E. coli O157:H7 strains may aide in the identification of gene clusters that contribute to its prevalence within the GIT of cattle [54].

In conclusion, significant differences in the composition of fecal microbiota of E. coli O157:H7 SS and NS cattle were identified, highlighting (i) a more diverse microflora in SS animals and (ii) specific bacterial OTUs that may be associated with the super-shedding state in cattle. These findings strongly support an important link between the intestinal microbial community and the density of this foodborne pathogen in cattle. However, whether E. coli O157:H7 overgrowth in super-shedders is promoted by intestinal dysbiosis or leads to bacterial population shifts remains unclear. Based on our results, future work could attempt to answer this question.

Supporting Information

Figure S1 Sequence data summary. Left: Relationship between processed sequence length (cleaned) and the number of sequences. Middle: Comparison of aligned sequence length and sequence length frequency. Right: The number of pyrotags for each library in super-shedding and non-shedding animals.

Figure S2 A Non-metric multidimensional scaling (NMDS) plot generated using fecal samples from different animal species. Cattle: denotes 22 cattle fecal samples from this study. The other metagenomic data were downloaded from http://trace.ncbi.nlm.nih.gov/Traces/sra/

Table S1 Taxa at five taxonomic levels (phylum to genus) for all samples combined. Total number of sequences for each taxa are shown in table.

Table S2 List of 72 differentially abundant OTUs.

Table S3 The number of OTUs shared by fecal samples as determined by the various clustering methods.

Acknowledgments

Special thanks are extended to staff at Highway 52 feeders for providing access to the steers for sampling. The technical assistance of S. Cook and S. Garland are also gratefully appreciated.

Author Contributions

Conceived and designed the experiments: TAM LBS. Performed the experiments: EDB LS RB KM LBS. Analyzed the data: YX EDB RZ. Contributed reagents/materials/analysis tools: TAM LBS. Wrote the paper: EDB YX TAM LBS.

References

1. Boyce TG, Sverdrup DL, Griffin PM (1995) Current Concepts - Escherichia coli O157:H7 and the hemolytic-uremic syndrome. New Engl J Med 333: 364–368.
2. Gould LH, Demma L, Jones TF, Hurd S, Vugia DJ, et al. (2009) Hemolytic-uremic syndrome and death in persons with Escherichia coli O157:H7 infection; foodborne diseases active surveillance network sites, 2000-2006. Clin Infect Dis 49: 1410–1415.
3. Graule IJ, Kaufla IT, Yoon JW, Hunt CW, Williams CJ, et al. (2002) Gastrointestinal tract location of Escherichia coli O157:H7 in ruminants. Appl Environ Microbiol 68: 2269–2277.
4. Naylor SW, Love JC, Besser TE, Mahajan A, Gunn GJ, et al. (2003) Lymphoid follicle-dense mucosa at the terminal rectum is the principal site of colonization of enterohemorrhagic Escherichia coli O157:H7 in the bovine host. Infect Immun 71: 1505–1512.
5. Matthews L, Low JC, Gally DL, Pearce MC, Moller DJ, et al. (2006) Heterogeneous shedding of Escherichia coli O157 in cattle and its implications for control. Proc Natl Acad Sci U S A: 547–552.
6. Omiakin F, MacRae M, Ogden ID, Strachan NJC (2003) Concentration and prevalence of Escherichia coli O157 in cattle feces at slaughter. Appl Environ Microbiol 69: 2444–2447.
7. Callaway TR, Carr MA, Edrington TS, Anderson RC, Nisbet DJ (2009) Diet, Escherichia coli O157:H7, and cattle: a review after 10 years. Curr Issues Mol Biol 11: 67–79.
8. Duncan SH, Louis P, Flint HJ (2004) Lactate-utilizing bacteria, isolated from human feces, that produce butyrate as a major fermentation product. Appl Environ Microbiol 70: 5810–5817.
9. Flint HJ, Duncan SH, Scott KP, Louis P (2007) Interactions and competition within the microbial community of the human colon: links between diet and health. Environ Microbiol 9: 1101–1111.
10. Guarner F, Malagelada JR (2003) Gut flora in health and disease. Lancet 361: 512–519.
11. Durro LM, Harzay GP, Smith TP, Bono JL, DeSantis TZ, et al. (2010) Animal-to-animal variation in fecal microbial diversity among beef cattle. Appl Environ Microbiol 76: 4858–4862.

12. Shanko OC, Kelty CA, Archibeque S, Jenkins M, Newton RJ, et al. (2011) Community structures of fecal bacteria in cattle from different animal feeding operations. Appl Environ Microbiol 77: 2992–3001.

13. Dowell SE, Callaway TR, Wolecott RD, Sun Y, McKeever T, et al. (2008) Evaluation of the bacterial diversity in the feces of cattle using 16S rDNA barcoded tag-encoded FLX amplicon pyrosequencing (STEPSAP). BMC Microbiol 8.

14. Care CC a (1993) Guide to the care and use of experimental animals. 3rd ed. Cambridge: MIT Press.

15. Gannon VPJ, Dsouza S, Graham T, King RK, Rahn K, et al. (1997) Use of the ribosomal database project: improved alignments and new tools for rRNA analysis. Nucleic Acids Res 25: 332–335.

16. Edgar RC (2010) Search and clustering orders of magnitude faster than BLAST. Bioinformatics 26: 2460–2461.

17. Cole JR, Wang Q, Cardenas E, Fish J, Chai B, et al. (2009) The Ribosomal Database Project: improved alignments and new tools for rRNA analysis. Nucleic Acids Res 37: D141–D145.

18. McDonald D, Price MN, Goodrich J, Navarro EP, DeSantis TZ, et al. (2011) An improved Greengenes taxonomy with explicit ranks for ecological and evolutionary analyses of bacteria and archaea. The ISME J 6: 610–616.

19. Kembel SW, Cowan PD, Helmus MR, Cornwell WK, Morlon H, et al. (2009) Introductory tutorial: phylogeny-independent, community-supported software for describing and comparing microbial communities. Appl Environ Microbiol 75: 7537–7541.

20. Sun YJ, Cai YP, Huse SM, Knight R, Farmerie WG, et al. (2012) A large-scale randomized controlled trial to assess the impact of dietary energy sources, feed characteristics, and feedlot management practices on the biodiversity of ruminal bacterial communities. J Anim Sci 87: 4151–4160.

21. Sun YJ, Cai YP, Huse SM, Knight R, Farmerie WG, et al. (2012) A large-scale randomized controlled trial to assess the impact of dietary energy sources, feed characteristics, and feedlot management practices on the biodiversity of ruminal bacterial communities. J Anim Sci 87: 4151–4160.

22. Schloss PD, Westcott SL, Ryabin T, Hall JR, Hartmann M, et al. (2009) Introducing mothur: open-source, platform-independent, community-supported software for describing and comparing microbial communities. Appl Environ Microbiol 75: 7537–7541.

23. Schloss PD, Gevers D, Westcott SL (2011) Reducing the effects of PCR amplification and sequencing artifacts on 16S rRNA-based studies. PLoS One 6: e25740.

24. Needleman SB, Wunsch CD (1970) A general method applicable to search for similarities in amino acid sequences of 2 proteins. J Mol Biol 48: 443–453.

25. Edgar RC, Haas BJ, Clemente JC, Quince C, Knight R (2011) UCHIME improves sensitivity and speed of chimera detection. Bioinformatics 27: 2194–2200.

26. Edgar RC, Haas BJ, Clemente JC, Quince C, Knight R (2011) UCHIME improves sensitivity and speed of chimera detection. Bioinformatics 27: 2194–2200.

27. Durso LM, Wells JE, Harhay GP, Rice WC, Kuehn L, et al. (2012) Comparison of 16S rRNA pyrosequences in quasilinear computational time. Nucleic Acids Res 39: e95.

28. Durso LM, Harhay GP, Smith TPL, Bono JL, DeSantis TZ, et al. (2011) Animal-to-animal variation in fecal microbial diversity among beef cattle. Appl Environ Microbiol 76: 4858–4862.

29. Shanko OC, Kelty CA, Archibeque S, Jenkins M, Newton RJ, et al. (2011) Community structures of fecal bacteria in cattle from different animal feeding operations. Appl Environ Microbiol 77: 2992–3001.

30. Dowell SE, Callaway TR, Wolecott RD, Sun Y, McKeever T, et al. (2008) Evaluation of the bacterial diversity in the feces of cattle using 16S rDNA barcoded tag-encoded FLX amplicon pyrosequencing (STEPSAP). BMC Microbiol 8.

31. Care CC a (1993) Guide to the care and use of experimental animals. 3rd ed. Cambridge: MIT Press.

32. Gannon VPJ, Dsouza S, Graham T, King RK, Rahn K, et al. (1997) Use of the ribosomal database project: improved alignments and new tools for rRNA analysis. Nucleic Acids Res 25: 332–335.

33. Kembel SW, Cowan PD, Helmus MR, Cornwell WK, Morlon H, et al. (2009) Introductory tutorial: phylogeny-independent, community-supported software for describing and comparing microbial communities. Appl Environ Microbiol 75: 7537–7541.

34. Schloss PD, Gevers D, Westcott SL (2011) Reducing the effects of PCR amplification and sequencing artifacts on 16S rRNA-based studies. PLoS One 6: e25740.