Effects of a Moderate or Aggressive Implant Strategy on the Rumen Microbiome and Metabolome in Steers

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The effects of growth-promoting implants have been well-defined for their ability to impact growth performance in beef cattle. Production-relevant microbes and microbiomes in the rumen have also been associated with growth traits. However, the role of implants on the rumen microbiome has not been determined. The objective of this study was to determine if different doses of implant hormones cause gain-associated ruminal microbial community changes. To assess this, a completely randomized design was used and 336 fall-born steers 450 to 470 days of age from the germplasm evaluation population at the US Meat Animal Research Center (Clay Center, NE) were divided into two treatment groups: 1) a moderate implant strategy (n = 167) of Revalor-IS (80 mg trenbolone acetate and 16 mg estradiol) followed by Revalor-S (120 mg trenbolone acetate and 24 mg estradiol) or 2) an aggressive implant strategy (n = 169) of Revalor-IS followed by Revalor-200 (200 mg trenbolone acetate and 20 mg estradiol). Steers were fed the same diet (57.0% dry-rolled corn, 30% wet distiller’s grains with solubles, 8.0% alfalfa hay, 4.25% vitamin and mineral supplement, and 0.75% urea, on a DM basis). On d 85 after implants administration, rumen contents were collected via orogastric tubing. Samples were sequenced to target and identify bacteria, archaea, and protozoa. Untargeted metabolomics was performed on rumen content using ultra high performance liquid chromatography high resolution mass spectrometry. Production data between implant strategies was analyzed using a mixed model ANOVA (SASv9.4, Cary, NC) followed by separation of least squares means. Microbial diversity between strategies did not differ for archaea or protozoa (P > 0.05). Average daily gain was different (P = 0.01; 1.72 vs 1.66 ± 0.02 kg, aggressive vs moderate, respectively); however, large microbial community shifts were not associated with implant strategy. Two metabolites, N-acetyllysine and N-acetylornithine, were found in greater abundance in the moderate implant strategy (P ≤ 0.04). Understanding associations between the rumen microbiome and implant strategies may allow improvement of growth efficiency in beef cattle.

Keywords: implants, microbiome, rumen, metabolomics, cattle
INTRODUCTION

The effects of growth-promoting implants have been well defined for their improvement of growth performance in beef cattle. These implants not only increase average daily gain (ADG), but also can also increase feed efficiency. Implants have been demonstrated to increase ADG by 30% and feed efficiency by 15%, ultimately serving as one of the most effective growth enhancing technologies available to feedlot operations (Duckett et al., 1996; Preston, 1999). In addition, implant strategies can be categorized by their ability to affect growth through the amounts and types of growth-promoting steroid hormones present within the implants. These hormones consist of natural or synthetic versions of estradiol, progesterone, or testosterone that synergistically work to augment weight gain (Parr, 2020).

Growth-promoting implant strategies often involve re-implanting cattle, especially ones on feed for more than 200 days. Using multiple implants allows producers to capture an increase in gain between 5 to 20%, making these strategies critical to feedlot operation profitability (Preston, 1999).

While multiple implant brands are available on the market, varying in potency of estrogenic and/or androgenic compounds, Revalor implants primarily consist of a synthetic testosterone, trenbolone acetate (TBA), and estradiol (E2). Revalor terminal implants primarily consist of a synthetic testosterone, varying in potency of estrogenic and/or androgenic compounds, to feedlot operations (Duckett et al., 1996; Preston, 1999). In addition, implant strategies can be implemented in various stages of production-relevant phenotypes (Artegoitia et al., 2017; Clemmons et al., 2020; Foroutan et al., 2020). Elucidating the interaction between the rumen microbiome and metabolome can improve understanding of ruminal microbiome and host interactions.

With the increased demand for sustainably-produced beef, there is a critical need for an improvement in feed efficiency and growth performance in beef cattle. The presence of estrogenic hormone receptors on different tissues, namely the gastrointestinal tract, may indicate the influence of these growth-promoting hormones on the rumen microbial communities, indirectly or synergistically improving gain and potentially altering ruminal metabolic activity. The objective of this study was to examine the rumen microbial communities and their metabolic features impacted by two growth-promoting implant strategies that may be influencing feed efficiency and growth in beef cattle.

METHODS AND MATERIALS

Animal Ethics Statement

The U.S. Meat Animal Research Center (MARC) Animal Care and Use Committee approved procedures utilized on animals in this study.

Experimental Design

Using a completely randomized design, germplasm evaluation composite breed steers 300-350 days of age from U.S. MARC (n = 336) were placed into feedlot pens. These steers consist of the 18 breeds in the U.S. with the highest registered animal numbers from breed association data. Steers were given ad libitum access to water and gradually adjusted to a feedlot diet consisting of 87% concentrate with vitamin and mineral supplements containing Rumensin and Tylan (Table 1). Unshrunk body weight was measured once a month throughout the entire study with an average initial body weights of 439.8 ± 43.1 kg.

A moderate or an aggressive implant strategy treatment was randomly assigned to steers and balanced across breed. The moderate implant strategy (n = 167) consisted of an initial

specifically noting 17β-estradiol degradation via the ruminal microbial communities (Isabelle et al., 2011). The interaction between the microbial communities and steroid hormones could influence shifts in the microbiome that leading to alterations in growth and metabolic activity.

The rumen microbiome has been studied in depth to demonstrate the importance of microbial communities on feed efficiency and gain in beef cattle (Myer et al., 2015). Previous studies have primarily focused on bacterial communities because of ease of technology and prevalence of bacterial communities in the gut. However, archaeal and protozoal communities also play important roles, namely in methane production and nitrogen cycling in the rumen which contribute to feed efficiency and growth (Hegarty et al., 2007; Li et al., 2017). Furthermore, investigations of metabolic activity are being performed to understand relationships between microbial communities and production-relevant phenotypes (Artegoitia et al., 2017; Clemmons et al., 2020; Foroutan et al., 2020). Elucidating the interaction between the rumen microbiome and metabolome can improve understanding of ruminal microbiome and host interactions.

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implant of Revalor-IS (80 mg trenbolone acetate and 16 mg estradiol) followed by the re-implantation treatment of Revalor-S (120 mg trenbolone acetate and 24 mg estradiol), which is an all-around moderate implant strategy suggested by Merck Animal Health (Kenilworth, NJ, USA). The aggressive implant strategy (n = 169) consisted of an initial implant of Revalor-IS, followed by the re-implantation treatment of Revalor-200 (200 mg trenbolone acetate and 20 mg estradiol), per suggestions by Merck Animal Health. The initial implant was administered 60 days after adjustment to the high-concentrate diet, with the re-implantation of the treatment implants administered 90 days after the initial implant of Revalor-IS.

After 85 days after the terminal implant, 200 mL of rumen content was sampled from each animal via orogastric tubing. A 1/2-inch tube was connected to a vacuum flask and vacuum pumped to sample ruminal liquid content with fibrous particles. Tubes were discarded between animals to prevent cross contamination. Ruminal content from individual animals was separated into 50 mL conical tubes and snap-frozen in liquid nitrogen. Samples were then stored at -80°C until further analysis.

DNA Extraction, Amplification, and Purification of Ruminal Content

Rumen samples containing fiber and content were sent to the University of Tennessee, Knoxville, TN for DNA extraction and purification, following the rumen digesta extraction protocol previously described by (Yu and Morrison, 2004). Following extraction and purification, the resultant DNA was then used for 16S and 18S rRNA amplicon sequencing, undergoing polymerase chain reaction (PCR) library construction. Primers for DNA amplification are listed in Table 2. Amplicon libraries of the 16S rRNA gene for bacteria (V1-V3) were prepared as described (Kozich et al., 2013), using primers 27F (Stahl, 1991) and 519R (Lane et al., 1985). As used in (Abbas et al., 2020), each 20 µL PCR amplification reaction contained 0.5 µL Terra PCR Direct Polymerase Mix (0.625 Units), 7.5 µL nucleae-free sterile water, 10 µL 2X Terra PCR Direct Buffer, 1 µL indexed fusion primers (10 µM), and 1 µL DNA (20 to 70 ng). Thermocycling conditions were as follows: initial denaturation at 98°C for 3 minutes, 25 cycles at 98°C for 30 seconds, 55°C for 30 seconds, and 68°C for 45 seconds, and a final extension of 68°C for 4 minutes.

For archaeal communities, amplicon libraries of the 16S rRNA gene for archaea (V3-V4) were prepared using primers ARC344F_TS (Wemheuer et al., 2012) and ARCH806R_TS (Takai and Horiokshi, 2000), found in Table 2. Each 10 µL run contained 4.2 µL nucleae-free sterile water, 1.25 µL 10X Buffer, 0.65 µL 25 mM MgCl2, 0.2 µL 25 mM dNTPs, 0.5 µL FC1:TS1 primer (4 µM), 0.2 µL HotStar Taq (QIAGEN, Valencia, CA, USA), 1 µL of bacterial indexes (1.875 µM), 1 µL of methanogen primers (1 µM), and 1 µL of DNA (15 ng/µL). Thermocycling conditions were as follows: 95°C for 15 minutes, 8 cycles of 94°C for 1 minute, 55°C for 30 seconds, and 72°C for 45 seconds, 24 cycles of 94°C for 20 seconds, 62°C for 30 seconds, and 72°C for 30 seconds, and a final extension of 72°C for 5 minutes.

Amplicon libraries of the 18S rRNA gene for protozoa (V3-V4) were prepared using primers P-SSU-316F (Sylvester et al., 2004) and GIC758R (Ishaq and Wright, 2014) (Table 2). Each 10 µL reaction consisted of 3.2 µL nucleae-free sterile water, 1 µL 10X Buffer, 0.65 µL 25 mM MgCl2, 0.2 µL 25 mM dNTPs, 0.5 µL protozoal primers (0.5 µM), 0.5 µL FC1:TS1 primer (4 µM), 0.2 µL HotStar Taq (QIAGEN, Valencia, CA, USA), 1 µL barcode, and 1 µL DNA (15 ng/µL). Thermocycling conditions were as follows: 95°C for 15 minutes, 8 cycles of 94°C for 1 min, 50°C for 30 seconds, and 72°C for 45 seconds, and 24 cycles of 94°C for 20 seconds, 62°C for 30 seconds, and 72°C for 30 seconds, and a final extension of 72°C for 5 minutes.

The libraries were sequenced using the V2 500-cycle kit on an Illumina MiSeq (Illumina, San Diego, CA, USA).

Bioinformatic Analyses in R

The Illumina 2x250 fastq sequencing files for bacterial and archaeal communities were processed through an R pipeline as described by (Callahan et al., 2016b). Fastq Illumina files were entered into R and open-source R packages ‘phyloseq’ (McMurdie and Holmes, 2013) and ‘DADA2’ (Callahan et al., 2016a) were used to perform filtering, merging, and taxonomic

| TABLE 1 | Diet on a dry-matter basis. |
| --- | --- |
| **Feedstuff** | **% Inclusion of Feedstuff** |
| Dry-rolled corn | 57.0 |
| Wet distillers’ grains with solubles | 30.0 |
| Alfalfa hay | 8.0 |
| Vitamin and mineral supplement with Rumensin and Tylan<sup>1</sup> | 4.25 |
| Urea | 0.75 |

<sup>1</sup>The supplement was formulated to contain the following on a 100% DM-basis: 17.1% non-protein nitrogen, 15.46% calcium, 0.35% phosphorus, 4.27% salt, 1.75% sodium, 0.53% magnesium, 0.65% potassium, 0.17% sulfur, 0.002% cobalt, 0.044% copper, 0.0045% iodine, 0.097% manganese, 0.00084% selenium, 0.16% zinc, 0.039% thiamin, Vitamin A 187.39 KIU/kg, Vitamin D3 47.49 KIU/kg, Vitamin E 246.92 IU/kg, and 746.5 g/ton monensin.

| TABLE 2 | Primers for PCR amplification of bacterial, archaeal, and protozoal communities. |
| --- | --- | --- | --- |
| **Regions** | **Primers** | **Sequences** | **Source** |
| **Bacteria** | V1-V3 | 27F | KRGTTYGATYNTGGCTCAG | Stahl, 1991 |
|  |  | S19R | G吴RTACTGGGCGGCKGCTG |
| **Archaea** | V3-V4 | ARC344F_TS | ACGGGGYGCGAGGCGCGG |
|  |  | ARCH806R_TS | G雀GACTAVGGGTATCTAAT |
| **Protozoa** | V3-V4 | P-SSU-316F | GCTTTCGWTGTTAGTATT |
|  |  | GIC758R | CAACGTGTCCTATKAAYCG | Lane et al., 1985 |
|  |  |  | Wemheuer et al., 2012 |
|  |  |  | Sylvester et al., 2004 |
|  |  |  | Ishaq and Wright, 2014 |
assignment. Forward and reverse reads were trimmed based on a quality score of 25 or greater. The expected errors per read were set at two for forward reads and four for reverse reads, thus filtering out any data not meeting these criteria (Edgar and Flyvbjerg, 2015). Following trimming and filtering of reads, ‘DADA2’ was used to identify real variants and correct for amplicon errors using a naïve Bayesian classifier (Wang et al., 2007; Callahan et al., 2016a). From DADA2, amplicon sequence variants (ASVs) were generated, which can moderately increase genetic resolution in comparison to 97% operational taxonomic units (OTUs) (Callahan et al., 2017; Glassman and Martiny, 2018). Quality-filtered forward and reverse reads were then merged and chimeras were removed. Sequences underwent taxonomic assignment to the genus level using the SILVA v132 database (Quast et al., 2012). The chimeras, Cyanobacteria, and Eukaryota were removed from the data.

Protozoal fastq sequencing files were also processed through the same pipeline as described by (Callahan et al., 2016b) and utilized the same open-source R packages. Forward and reverse reads were trimmed based on quality score (Q ≥ 25). The expected errors per read were set at two for forward and five for reverse reads, thus, filtering out data not meeting these criteria. Samples with initially less than 10,000 input reads were removed. Then, DADA2 was used to identify ASVs, forward and reserve reads were merged, and chimeras were removed. Sequences underwent taxonomic assignment to the genus level using the SILVA v132 database, and singletons, Cyanobacteria, and any communities within the kingdom Bacteria were removed from the data.

Rumen Metabolite Extraction and Analysis

After removing from storage in a -80°C freezer, a 2 mL aliquot of rumen content was transferred to two sterile 2 mL microcentrifuge tubes and centrifuged at 6,000 x g for 15 minutes at 4°C. Supernatant was aspirated and combined from each tube using a 3 mL syringe and then filtered through a 0.22 μM syringe filter (MidSci, St. Louis, MO, USA) to a sterile 2 mL cryovial. Samples were stored at -20°C until analyzed. Samples were processed at the Biological and Small Molecule Mass Spectrometry Core (BSMMC), University of Tennessee, Knoxville, TN (RRID: SCR_021368). Water soluble metabolites were extracted from 100 μL rumen content using 4:4:2 acetonitrile/methanol/water with 0.1 M formic acid as previously described by (Domingo-Almenara et al., 2019). Following extraction, the supernatant was dried under N2 then resuspended in 300 μL water. All solvents used were HPLC grade and purchased from Fisher Scientific (Hampton, NH, USA). Metabolites were identified using ultra high performance liquid chromatography high resolution mass spectrometry (UHPLC-HRMS) (Thermo Scientific, San Jose, CA, USA) with a previously validated untargeted metabolomics method by (Lu et al., 2010). Prior to mass analysis, reverse-phase (RP), ion-pairing chromatography was used to separate metabolites based on polarity. A Synergi Hydro RP column (100 mm x 2.1 mm, 2.6 μm, 100 Å) and an UltiMate 3000 pump (Thermo Fisher) were used for chromatographic separation. An Exactive Plus Orbitrap MS (Thermo Fisher) was coupled to the UHPLC system and used for mass analysis.

Statistical Analyses

For bacterial, archaeal, and protozoal sequences, alpha- and beta-diversity metrics were analyzed in R. For alpha diversity, observed communities were calculated along with expected species (Chao1) as well as richness and diversity estimates were performed (Shannon). Alpha diversity was calculated individually for bacteria, archaea, and protozoa, and for each alpha diversity metric, differences between treatments were tested in SAS v9.4 (SAS Institute, Cary, NC, USA) using mixed model analysis of variance and Fisher’s least significant difference mean comparisons. Beta diversity was measured with a Bray-Curtis distance matrix producing a principal coordinates analysis (PCoA). Following, a PERMANOVA was conducted with 999 permutations to determine significance of the Bray-Curtis PCoA using ‘vegan’ in R (Oksanen et al., 2007). Differential abundances between treatments for bacterial, archaeal, and protozoal communities were individually calculated using R package ‘DESeq2’ (Anders and Huber, 2010; Love et al., 2014).

To visualize raw UHPLC-HRMS data and identify rumen metabolites, an open source metabolomics software package with a grouping algorithm and peak alignment feature, Metabolic Analysis and Visualization Engine (MAVEN), was used. Metabolites were identified in MAVEN based on the exact mass and retention time of each metabolite (Clasquin et al., 2012). For all 115 identified metabolites, area under the curve was integrated and these data were used for further analysis. Data from MAVEN were processed and analyzed in MetaboAnalyst4.0 (Chong et al., 2018), with significant pathways identified using a global test with relative-betweenness centrality topology analysis with the reference library of Escherichia coli K-12 MG1655 (Kanehisa et al., 2013). Rumen metabolites were imported into SAS v9.4 (SAS Institute, Cary, NC, USA) and normality was assessed using the PROC UNIVARIATE procedure on raw, untransformed data. Normally distributed data were identified using Shapiro-Wilks (W) score of ≥ 0.85 as well as visual assessment of histograms and residuals. Non-normal data (W < 0.85) were then log-transformed and analyzed with a mixed model analysis of variance (ANOVA) to compare between moderate and aggressive implant strategies with a covariate of initial body weight. To address multiple testing, a Fisher’s-protected LSD was used. Significance was determined using P = 0.05.

To understand associations between low abundance metabolites and implant treatment, binary variables were created to indicate the presence of absence of low abundance metabolites in each sample. The frequency procedure (PROC FREQ; SAS v9.4) was used to compare the presence or absence of these metabolites by each treatment using Chi Square analyses (SAS Institute, Cary, NC, USA). When low-abundance metabolites were found significantly different between moderate and aggressive implant strategies, least significant differences were used to differentiate between the ADG of
moderate and aggressive implant strategies. Significance was determined using $P = 0.05$.

RESULTS

Steer Performance
Steers within the aggressive implant strategy had significantly greater ADG (1.72 ± 0.02 kg) compared to steers within the moderate implant strategy (1.66 ± 0.02 kg; $P < 0.02$).

Ruminal Bacterial Communities
Within ruminal bacterial communities, the SILVA v138 database identified 18,381 taxa within 6 taxonomic ranks. Chimeras were identified to represent 2% of sequences after filtering. Following taxonomic assignment, alpha- and beta-diversity metrics were measured to identify richness and diversity. Observed species richness of bacterial communities in the rumen was determined from samples, then estimated richness of communities was calculated for the Chao1 metric, and samples were analyzed for species richness and diversity using Shannon diversity indices (Figure 1). Richness and evenness of species (Shannon) were different between the moderate and aggressive implant strategies ($P < 0.01$), however Chao1 and observed species were not ($P > 0.05$). A PCoA identified any dissimilarity among clustered samples, using the Bray-Curtis distance method (Figure 2). Beta diversity was analyzed with a PERMANOVA for the Bray-Curtis PCoA with no dissimilarity found between treatments ($P > 0.05$).

There were no differences in relative abundance of any bacterial taxa between treatments. The majority of bacterial phyla identified in rumen content were Bacteroidota (52.79%) and Firmicutes (36.48%), as shown in Figure 3. The Proteobacteria accounted for 9.28% of reads. Many phyla represented less than 1% of relative abundance. Top 10 bacterial phyla present in rumen of cattle within both treatments are presented in Figure 3. Identified bacterial genera (Figure 4) consisted primarily of Prevotella, making up 38.28% of samples, but also Succinivibrionaceae UCG-001 (8.50%), Dialister (7.39%), Mitsuokella (2.42%), and others under 2% abundance. Presence of other genera were less than 1% of relative abundances in rumen fluid.

Ruminal Archaeal Communities
For archaeal communities, the SILVA v138 database identified 858 taxa within 6 taxonomic ranks. Chimeras were identified to represent 23.19% of sequences after filtering. Alpha-diversity metrics of observed archaeal abundances (Observed), expected

![Figure 1](https://example.com/figure1.png)  
**FIGURE 1** | Alpha-diversity metrics for bacterial communities with observed abundances of communities, expected abundances (Chao1), and richness and diversity measured with Shannon diversity index. Aggressive treatments are represented in purple and moderate treatments are represented in yellow. Significant differences were only noted in the Shannon diversity index between treatments ($P < 0.05$).
species abundance (Chao1), and richness and evenness estimates (Shannon) were calculated and not different between treatments ($P > 0.05$; Figure 5). Beta diversity, measured through a Bray-Curtis PCoA, visually demonstrated similarities between treatment groups and the PERMANOVA performed showed no differences between the moderate and aggressive implant strategies ($P > 0.05$; Figure 6).

There were no differences between the moderate and aggressive implant strategies at any taxonomic level after performing analyses in DESeq2. The majority of archaea at the phylum level were Eurarchaeota, making up 99% of samples. At the genus level (Figure 7), the majority of archaeal communities identified were Methanobrevibacter (91.75%). Minor archaeal communities consisted of the genera Methanosphaera (3.81%) and Candidatus methanomethylophilus (2.82%).

**Ruminal Protozoal Communities**

After filtering and processing in R for protozoal communities, the SILVA v132 database identified 220 taxa within 6 taxonomic ranks. Chimeras were identified to represent 4.9% of sequences after filtering. Out of the 336 samples, 128 protozoal fastq files did not meet filtering standards and were removed from data, resulting in 208 final samples for the protozoal communities (moderate = 109, aggressive = 99). After taxonomic assignment, 220 taxa were identified. Alpha diversity was measured for observed abundance, estimated abundance (Chao1), and species evenness and richness (Shannon) (Figure 8) and no differences were found between treatments ($P > 0.05$). Beta
diversity was measured with a Bray-Curtis PCoA (Figure 9) and after testing with a PERMANOVA, was not significant between the moderate or aggressive strategies ($P > 0.05$).

At the phylum level, SAR was the most abundant category, making up 99% of samples. Top 10 phyla relative abundance are shown in Figure 10. At the genus level (Figure 10), the majority of protozoal communities identified were Entodinium (58.86%). Minor protozoal communities consisted of Ophryoscolex (6.8%), Isotricha (3.2%) and Polyplastron (1.6%). There was a difference between the relative abundance of Isotricha between the moderate (3.0% ± 0.03) and aggressive (3.3% ± 0.03) implant strategies ($P < 0.05$).

**Rumen Metabolite Data**

After untargeted metabolomics were performed on rumen samples, 115 unique metabolites were identified. Partial least-squares discriminant analysis of the 115 metabolites demonstrated similarity in the metabolic profile of the moderate and aggressive implant strategies (Figure 11). Between the moderate and aggressive implant strategies, two metabolites were found to be different (Table 3). In the moderate treatment group, N-acetyllysine ($P ≤ 0.03$) and N-acetylnornithine ($P ≤ 0.04$) were greater than the aggressive treatment. The covariate of initial body weight was not significant in any model ($P > 0.05$). From analysis in MetaboAnalyst 4.0, these metabolites are most similarly correlated to arginine synthesis in metabolic pathways ($P < 0.05$).

**DISCUSSION**

Improving the efficiency of weight gain and growth performance in beef cattle is one way to contribute to a reliable food source to feed a growing population. Growth-promoting implant strategies remain one of the most effective ways to improve gain in beef cattle, increasing ADG up to 30% and feed efficiency by 15% (Preston, 1999). The use of aggressive implant strategies is thought to improve ADG beyond the capabilities of moderate implant strategies due to increased concentrations of androgenic and estrogenic compounds and altered ratio of the hormones.

The use of growth-promoting hormones in management strategies may be influencing other factors that impact weight gain and feed efficiency in beef cattle, such as the rumen microbiome. The rumen microbiome has been previously identified to impact feed efficiency in beef cattle through the conversion of nutrients to energy for the host (Mizrahi, 2011; Myer et al., 2015). Further, fermentative products of the rumen microbial communities, particularly volatile fatty acids, supply energy to the host and play a role in metabolic pathways (Anantasook et al., 2013). While one alpha diversity...
measurement for bacterial communities did differ, the difference is likely not biologically relevant as no microbial communities differed between treatments. This is further supported by little differences seen in the other diversity metrics calculated when assessing the archaean and protozoal communities between treatments.

The few differences in the rumen microbiome and metabolome may be explained by all steers being fed the same diet. While these implant strategies are identified as moderate and aggressive, there was overall little biological differences in ADG, likely because the cattle were tubed for rumen fluid at equal days on feed, irrespective of 12th rib fat thickness. It is likely the influence of the composite breed type of the cattle would allow them to reach maturity at different times and were on varying places of the growth curve at the time of tubing for rumen fluid. A previous study conducted demonstrated a similar core rumen microbiome composition across animals in different climates and farming practices, with the major driving factor being diet (Henderson et al., 2015). Therefore, if the hormones present in the implant strategy are not influencing the rumen microbiome, animals receiving the same diet within the same environment would have rumen microbial profiles that are very similar.

Previous research has not identified estrogenic and androgenic receptors on the rumen epithelium of cattle with a fully developed rumen, potentially signifying the lack of a mechanism for these hormones to enter the rumen environment. While estrogen-like receptor-α, highly similar in sequence similarity to estrogen receptors (Giguère et al., 1988), has been researched on the rumen epithelium in calves 14-42 days of age (Connor et al., 2014), these studies have not identified other binding sites for estrogenic and androgenic compounds found in growth-promoting implants. Further, the estrogen-related receptors have not been identified to bind to estrogens or related steroid hormones, but function to increase transcription of genes in energy metabolism through an unknown endogenous ligand (Schreiber et al., 2003). Future studies should assess receptors present on rumen tissue that interact with growth-promoting hormones and the rumen environment.

The presence of estrogenic receptors on gastrointestinal tissues has been identified in human studies, prompting further research into the gut-brain axis involving the activity of estrogens. Decreases in the diversity in the gut microbiome in humans have been linked to a reduction of β-glucuronidase, which functions to deconjugate estrogen and phytoestrogens into
their active and circulating forms (Baker et al., 2017). If these estrogenic receptors can be identified on rumen tissue, there is potential for additional circulating estrogen to spur additional production of GH and IGF-I to increase muscle tissue growth through binding sites present on muscle tissue.

In bovines, androgenic compounds have been demonstrated to have direct and indirect effects on muscle and growth through binding directly to receptors in muscle tissues (Sauerwein and Meyer, 1989), while indirectly spurring production of GH and IGF-I (Grigsby and Trenkle, 1986). Bovine skeletal muscle directly interacts with free androgens that are introduced via implanting strategies through androgen receptors, which in addition to estrogen, can cause additive growth in muscle tissue (Snochowski et al., 1981). While GH and IGF-I play an important role in muscle development, they have not been associated with rumen functionality. Research is still needed to identify potential receptors in the rumen in order to draw conclusions.

While the protozoal genus *Isotricha* was found in greater presence in the moderate implant group, no other microbial community was found significantly different between treatments. Interestingly, past studies have examined the defaunation of protozoal communities to understand their impact on the rumen microbiome (Newbold et al., 2015). The protozoal communities present in the rumen often outweigh their positive interaction with digestion of fiber and starch particles (Mendoza et al., 1993; Takenaka et al., 2004) by decreasing efficient N utilization in the rumen, thus decreasing energy...
provided to the host. Further, ruminal protozoa function to phagocytize some bacterial communities, which decreases availability of microbial protein (Ushida et al., 1990). When investigating impacts of saponins on methane production and the composition of the rumen microbiome, *Isotricha* were found in greater abundance in animals with higher CH₄ production (Tan et al., 2020). Generally, a higher methane production is associated with decreased feed efficiency in cattle (Hegarty et al., 2007), which may be indicative of rumen microbial composition (McGovern et al., 2020). This association may indicate that the steers within the moderate implant strategy are not using energy as efficiently as those in the aggressive implant strategy, which may play a role in altering gain. Despite the interesting functionality behind host energy supply, protein metabolism, and methane production from protozoal communities, the *Isotricha* populations cannot be directly associated with differences in the metabolome or gain within this study and influences can only be speculated.

Out of the 115 metabolites identified in rumen fluid, only two were significantly different between the moderate and aggressive implant strategies. While there were significant differences in metabolites, it is important to note that there were no significant differences between bacterial and archaeal communities in the rumen microbiome. Thus, the differences in metabolites likely does not stem from rumen bacterial or archaeal composition, although the metabolic activity of these communities may have been altered. With the stability of the bacterial and archaeal communities and only one protozoal genus differing between the moderate and aggressive implant strategy, significant changes in metabolite concentrations could be due to host metabolic activity rather than any microbial activity or fermentation. N-acetyllysine is an acetyl-derivative of lysine, an essential amino acid that is critical for growth in beef cattle. Interestingly, lysine functions as one of the most limiting amino acids in growing cattle (Richardson and Hatfield, 1978), as it contributes greatly to maximizing lean tissue deposition (Friesen et al., 1994). The acetylation of lysine may explain why the moderate implant treatment resulted in lower ADG, as less lysine may be contributing towards muscle growth. However, this is likely not the sole explanation for the decrease in ADG in the moderate implant treatment group, as amounts of androgenic and estrogenic hormones directly influence skeletal muscle by stimulating GH and IGF-I to bind to muscle, causing proliferation of muscle satellite cells and resulting in additional growth (Froesch and Zapf, 1985). Based on this, the use of growth-promoting implants appears unlikely to influence the

**FIGURE 8** | Alpha-diversity metrics for protozoal communities with observed abundances of communities, expected abundances (Chao1), and richness and diversity measured with Shannon diversity index. There were no significant differences between alpha-diversity metrics between moderate and aggressive implant strategies. Aggressive treatments are represented in purple and moderate treatments are represented in yellow.
rumen metabolome, rather directly acting on muscle tissue to impact ADG.

While previous studies have analyzed rumen metabolites (Artegoitia et al., 2017; Clemmons et al., 2020), the pathways and activity of many of these metabolites play multiple functions throughout the rumen and host metabolism, and thus can be difficult to identify true functionality. In addition, other studies have concluded that the origin of the metabolites can be difficult to trace, with half of the rumen metabolites coming from bovine origin and the remaining from microbial origin (Saleem et al., 2013). Furthermore, previous studies have also found a wide variation among the abundance of metabolites, leading to belief that more research is required to be able to associate rumen metabolites with diet and feed efficiency (Saleem et al., 2013). While the presence of N-acetylysine may be due to shifts in protozoal communities, other metabolites such as N-acetylornithine could be due to differential absorption across the rumen epithelium. The lack of differences in ruminal bacterial and archaeal communities may demonstrate that the metabolites found in this study are of host origin. Notably, there are few studies involving N-acetylornithine and its metabolic effect on beef cattle, making associations more difficult. Previous studies have suggested the relationship of N-acetylornithine with enzymatic activity for arginine and lysine biosynthesis (Ledwidge and Blanchard, 1999), may be contributing towards protein use in cattle.

Previous studies analyzing how shifts in protozoal communities influence rumen metabolites have commonly
identified purine derivatives due to the effects of purine derivatives on microbial protein synthesis (Chen and Gomes, 1992). Interestingly, the metabolites identified in this study that could be related to significant shifts in protozoal communities were not identified in other studies. Purine derivatives, such as allantoic acid and hypoxanthine, were identified (Chen and Gomes, 1992); however, they did not significantly differ between treatments. In addition, metabolites that have been previously identified to be influenced by protozoa in lambs such as aminoacidipate, cholate, and pantothentic acid did not significantly differ in this study (Morgavi et al., 2015). Aminoacidipate acid has been identified to be influenced by protozoal communities, and plays an important role in lysine metabolism. While aminoacidipate acid was not different between treatments, N-acetylysidine also affects presence of lysine in biological fluids and was found different between treatments.

Understanding the relationship of microbial communities with host phenotypes for feed efficiency and growth may allow for better identification of efficient beef cattle. The topic of growth-promoting implants has been well-reviewed, identifying that implants greatly improve ADG in beef cattle and making implants one of the most effective management strategies in increasing growth (Duckett et al., 1996; Preston, 1999). Additionally, studies have identified the importance of the rumen microbiome in nutrient conversion and overall feed efficiency (Myer et al., 2015; Auffret et al., 2020). With rumen microbial communities producing fermentative products and metabolites that may be linked to phenotypic traits, metabolites should be further characterized in the rumen.

Table 3 | Metabolites differing between aggressive and moderate implant strategies.

| Metabolite      | Aggressive Treatment | Moderate Treatment | P-value * |
|-----------------|----------------------|--------------------|-----------|
| N-acetillysine  | 78,453 ± 5606        | 98,259 ± 5606      | 0.02      |
| N-acetylornithine | 96,563 ± 5006        | 110,899 ± 5006     | 0.03      |

*Significance determined at P ≤ 0.05.

Fig. 11 | Partial least-squares discriminant analysis of 115 rumen fluid metabolites identified through untargeted metabolomics. The aggressive implant strategy is represented in yellow and the moderate implant strategy is represented in blue. Significant overlap of metabolites present in moderate and aggressive implant strategies illustrates similarity of rumen fluid metabolic profiles. Shaded regions represent a 95% confidence interval around the metabolites per treatment strategy.

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Conclusions

There were no differences in the ruminal bacterial and archaeal community abundance between the moderate and aggressive implant strategies. Between a moderate and aggressive implant strategy, only one protozoal genus differed, potentially indicating that the growth-promoting hormones included in implants have minimal influence on the rumen. While no differences were identified between bacterial or archaeal communities, this study corroborated previously determined differences in ADG due to implant strategy, while finding one ruminal protozoal community that differed in steers given a moderate or aggressive implant strategy. Future directions with this study could involve multiple brands of implant strategies as well as including non-implanted cattle to determine more ruminal and metabolic differences.

Data Availability Statement

The datasets generated and analyzed for this study can be found in the Sequence Read Archive (PRJNA811160) [https://www.ncbi.nlm.nih.gov/bioproject/PRJNA811160/].

Ethics Statement

The animal study was reviewed and approved by the U.S. Meat Animal Research Center (MARC) Animal Care and Use Committee, USDA-ARS-U.S. Meat Animal Research Center, Clay Center, NE.

Author Contributions

JW, KH, AL-P, HF, LK, and PM contributed to conception and design of the study. MH performed all sample and statistical analyses. MH conducted data analyses and CC conducted metabolomics analyses. MH, PM, KM, and LS aided in the interpretation of results. MH wrote the first draft of the manuscript. CC, SC, AL-P, LS wrote sections of the manuscript.
All authors contributed to manuscript revision, read, and approved the submitted version.

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**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fanim.2022.889817/full#supplementary-material

**Supplementary File 1 | Genus-level abundances for ruminal bacterial communities.**

**Supplementary File 2 | Genus-level abundances for ruminal archaeal communities.**

**Supplementary File 3 | Genus-level abundances for ruminal protozoal communities.**

**Supplementary File 4 | Raw data for rumen metabolomics.**
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