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Assessment of Selected Antibiotic Resistances in Ungrazed Native Nebraska Prairie Soils

Lisa M. Durso,* David A. Wedin, John E. Gilley, Daniel N. Miller, and David B. Marx

Abstract

The inherent spatial heterogeneity and complexity of antibiotic-resistant bacteria and antibiotic resistance (AR) genes in manure-affected soils makes it difficult to sort out resistance that can be attributed to human antibiotic use from resistance that occurs naturally in the soil. This study characterizes native Nebraska prairie soils that have not been affected by human or food-animal waste products to provide data on background levels of resistance in southeastern Nebraskan soils. Soil samples were collected from 20 sites enumerated on tetracycline and cefotaxime media; screened for tetracycline-, sulfonamide-, β -lactamase-, and macrolide-resistance genes; and characterized for soil physical and chemical parameters. All prairies contained tetracycline- and cefotaxime-resistant bacteria, and 48% of isolates collected were resistant to two or more antibiotics. Most (98%) of the soil samples and all 20 prairies had at least one tetracycline gene. Most frequently detected were *tet(D)*, *tet(A)*, *tet(O)*, *tet(L)*, and *tet(B)*. Sulfonamide genes, which are considered a marker of human or animal activity, were detected in 91% of the samples, despite the lack of human inputs at these sites. No correlations were found between either phenotypic or genotypic resistance and soil physical or chemical parameters. Heterogeneity was observed in AR within and between prairies. Therefore, multiple samples are necessary to overcome heterogeneity and to accurately assess AR. Conclusions regarding AR depend on the gene target measured. To determine the impacts of food-animal antibiotic use on resistance, it is essential that background and/or baseline levels be considered, and where appropriate subtracted out, when evaluating AR in agroecosystems.

Core Ideas

- Native Nebraska prairie soils have measurable amounts of antibiotic resistance.
- Phenotypic and genotypic measures of resistance vary within and between sites.
- Ungrazed prairie soils can provide background data on resistance in Nebraskan soils.
- Assessments of resistance on farms should include the collection of background data.
- Background resistance should be considered when measuring impact of management.

ANTIBIOTIC-RESISTANT bacteria and antibiotic resistance (AR) genes occur naturally in soil and water from around the globe (Allen et al., 2010; D'Costa et al., 2006; D'Costa et al., 2007; Durso et al., 2012; Cytryn, 2013). Recent studies have highlighted the potential for soil-borne antibiotic-resistant bacteria and AR genes to affect human clinical outcomes, and it has been proposed that many important AR genes originally came from bacteria living in soil (Fosberg et al., 2012; Wright, 2010). In addition to soils' hypothesized natural capacity to contribute to antibiotic-resistant disease in humans, soil is thought to serve as a sink and transport medium for resistance associated with human use of antibiotic drugs via land application of waste products from municipal wastewater, hospitals, pharmaceutical and industrial manufacturing, and agriculture (Ashbolt et al., 2013; Rutgersson et al., 2014; Koh et al., 2015; Amos et al., 2015; Williams-Nguyen et al., 2015).

Studies that characterize antibiotic-resistant bacteria and AR genes in animal manure-affected soils provide valuable information on what is currently present in a sample site. However, if background and baseline levels of AR are not collected, measurements taken in these manure-affected soils can result in conflated data. It is difficult to sort out the resistance that is a result of human antibiotic use from the resistance that is a result of the inherent spatial heterogeneity, complexity, and dynamics of antibiotic-resistant bacteria and AR genes in soils.

One way to start to address this problem is to characterize soils that have been minimally affected by human or food-animal waste products. These areas can be difficult to identify, especially when one requires documentation that specific land has not been grazed by cattle, sheep, goats, or other food animals in the past. In Nebraska, where the majority of land has been converted to agricultural uses, the main candidates for these kinds of background data collections are land that is used for cemeteries and land that is conserved as prairies (Fierer et al., 2013). In this study, soils were collected from native prairies where the land-owners could confirm no grazing by food animals (including no accidental grazing) for at least the last 20 yr.

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Abbreviations: agBMPs, agricultural best management practices; AR, antibiotic resistance; CLSI, Clinical Laboratory Standards Institute; PCR, polymerase chain reaction.

With over 49,000 farms and ranches covering 92% of the state's total land area, Nebraska is one of the nation's top agricultural states. Over 5 million cattle are fed and marketed each year on 4570 cattle feeding operations statewide (USDA National Statistics Service, 2012). Because of the large footprint of animal agriculture in the state, Nebraska is an appropriate place to investigate agriculturally affected AR in soils. Although the majority of the Nebraska land area is involved in agriculture, there is also a network of native prairie sites being preserved by groups such as The Wachiska Audubon Society and the Nebraska Games and Parks Commission in conjunction with private landowners. The native prairie sites have similar soil types to the surrounding agricultural land and have been exposed to similar macroecologic conditions (i.e., temperature, precipitation). As such, they can provide useful information on naturally occurring resistance in southeastern Nebraska, including estimates of resistance in the absence of agricultural inputs and estimates of how the measured parameters are distributed within and between locations.

There is broad agreement in both scientific and policy communities that AR in agroecosystems needs to be reduced. However, in the absence of data on the levels and types of resistance that occur in minimally affected settings, it is difficult to set realistic targets for reduction. Additionally, even when a specific drug type, bacteria, or gene is identified as a means to measure resistance, little data are available to determine the variability for the chosen parameter within a sample, site, or region.

The goal of this study was to characterize selected phenotypic and genotypic AR traits in native southeastern Nebraskan prairie soils. To minimize potential sources of variation, sites were selected based on geographic proximity, and all sample collections were performed within the same week. Our working hypothesis was that the background levels of resistance for

each of the measured parameters would be similar within and between the tested prairies.

Materials and Methods

Prairie Sites

Potential prairie sites were identified in collaboration with the University of Nebraska School of Natural Resources, The Wachiska Audubon Society, and the Nebraska Games and Parks Commission. Land owners were contacted individually to obtain information on land status and history, to verbally confirm that the land had not been purposefully or accidentally grazed in the last 20 yr, and to obtain permission to collect samples. Twenty sites in five counties were chosen for the study (Table 1; Supplemental Fig. S1). The counties that were sampled are located in the part of Nebraska that was originally categorized as upland tall grass prairie.

Soil Sampling

Soil sampling was conducted in July 2012. Within each prairie, five spatially separated sites were chosen for sampling. Each sample consisted of six 30-cm-deep cores obtained using a step-on soil probe (19.2 mm i.d.). Due to exceptionally dry or rocky conditions, samples from Prairies 3, 7, and 19 consisted of a total of 12 cores 15 cm deep, and Prairies 1 and 2 were a mix of full and half cores. For each site, all cores were placed in a whirl-pac bag and returned to the laboratory in a cooler with ice within 4 h of collection. At the laboratory, whirl-pac bags were removed from the cooler and homogenized by hand before removing aliquots for bacterial plating (resulting in bacterial isolates), soil physical and chemical analysis, and DNA isolation of bulk soil extracts (described below).

Table 1. Prairie sites.

Prairie ID	County	Area	Name	GPS coordinates
		ha		
1	Lancaster	23.9	9 Mile Prairie East (UNL)	40°52'0.88" N 96°48'29.5" W
2	Lancaster	64.6	9 Mile Prairie West (UNL)	40°52'4.71" N 96°49'8.8" W
3	Pawnee	4.1	Private A	private land†
4	Pawnee	5.4	Private B	private land
5	Pawnee	5.5	Private C	private land
6	Pawnee	4.5	Private D	private land
7	Richardson	3.8	Camp Cornhusker	40°00'0.9" N 95°56'0.1" W
8	Pawnee	15.4	Private E	private land
9	Otoe	5.2	Dieken (Wachiska Audubon)	40°40'03.8" N 96°17'26.7" W
10	Pawnee	6.4	Private F	private land
11	Pawnee	12.9	Private G	private land
12	Pawnee	4.1	Private H	private land
13	Pawnee	7.9	Private I	private land
14	Pawnee	6.4	Klapka (Wachiska Audubon)	40°08'33.59" N 96°04'02.48" W
15	Lancaster	30.5	Lincoln Airport Authority Meadow	40°52'19.72" N 96°48'21.53" W
16	Lancaster	9.6	Mitchell Prairie (UNL)	40°37'34.34" N 96°42'13.67" W
17	Johnson	3.9	Private J	private land
18	Lancaster	6.7	Prairie Pines (UNL)	40°50'34.61" N 96°33'35.25" W
19	Johnson	4.2	Table Rock State Wildlife Management Area	40°24'02.45" N 96°13'01.69" W
20	Pawnee	13.7	Wildcat (Wachiska Audubon)	40°15'10.38" N 96°10'43.85" W

† Location data for private land are available on request after consultation with private landowners.

Soil Physical and Chemical Analysis

Moisture content was determined for 5 g of soil by mass loss on drying at 105°C for 24 h. Additional physical and chemical parameters, including organic matter, potassium, sulfate, zinc, iron, manganese, copper, calcium, magnesium, sodium, nitrate, ammonium, Mehlich P, and excess lime, were measured by Ward Laboratories. Soil samples for these analyses were shipped on ice and analyzed within 48 h of collection.

Bacterial Culture and Characterization

After homogenization of the soil samples, 10 g of soil was weighed out into a fresh whirl-pack filter bag, 90 mL of sterile phosphate-buffered saline (ThermoFisher Scientific) was added, and bags were manually mixed. *Escherichia coli*, total coliforms, and *Enterococcus* were enumerated using Quantitray (IDEXX Laboratories). *Escherichia coli* and total coliform trays were incubated at 37°C for 24 to 28 h, and *Enterococcus* trays were incubated at 42°C for 24 h.

Phenotypic enumeration of selected ARs was performed on the 100 bulk soil samples, and isolates were picked from these plates for further characterization. For soil bacterial colony counts, soil samples ($n = 100$; five sites from each of 20 prairies) were plated onto R2A (Becton Dickinson), R2A with tetracycline (TR2A) ($16 \mu\text{g mL}^{-1}$), and R2A with cefotaxime (CR2A) ($4 \mu\text{g mL}^{-1}$) using an Eddy Jet spiral plater (Neutec Group). Tetracycline was chosen because it is commonly assayed in environmental samples. Cefotaxime was chosen as a representative of third-generation cephalosporins, which are used to treat hospital-acquired infections and invasive *Salmonella* infections in children (Fey et al., 2000). Where necessary to obtain statistically appropriate colony counts on the plates, samples were serially diluted before being spiral plated. Plates were incubated at 37°C for 18 to 24 h and enumerated using a standard spiral plate procedure (Gilchrist et al., 1973). Three isolates were picked for each sample from the plain R2A plates ($n = 300$ isolates), struck for isolation, grown overnight, and frozen at -80°C for later characterization. Disk diffusion assays were performed on isolates according to standard Clinical Laboratory Standards Institute (CLSI) methods using Mueller-Hinton broth (Becton-Dickinson) on 150-mm \times 15-mm plates (CLSI, 2012). The CLSI clinical breakpoints were used to assign isolates sensitive or resistant status, and isolates displaying intermediate levels of resistance were counted as sensitive. Twelve drugs were used in the disk diffusion assays: amoxicillin/clavulanic acid 20 μg , ampicillin 10 μg , cefoxitin 30 μg , ceftriaxone 30 μg , chloramphenicol 30 μg , ciprofloxacin 5 μg , gentamycin 10 μg , kanamycin 30 μg , nalidixic acid 30 μg , streptomycin 10 μg , sulfamethoxazole trimethoprim 25 μg , and tetracycline 30 μg .

Polymerase Chain Reaction

Polymerase chain reaction (PCR) detection of selected AR genes was performed on the 100 bulk soil samples. DNA isolations were performed on each of the bulk soil samples ($n = 100$) using the MoBio PowerSoil kit according to the manufacturer's directions, except that a Mini Beadbeater-8 (Biospec Products) was used for the cell lysis step. Qualitative tetracycline resistance gene assays were performed as previously

described (Ng et al., 2001) with the exception that Jumpstart RedTaq Master Mix (Sigma) was used. Four multiplex reactions were run, evaluating 14 tetracycline resistance genes. Positive control strains were created by cloning the target PCR fragment into TOP10 *E. coli* cells using a TOPO TA Cloning Kit for Subcloning, with One Shot TOP10 chemically competent *E. coli* cells (Life Technologies), and are available on request. Two sulfonamide resistance genes, *sul*(I) and *sul*(II), were assayed using primers described by Pei et al. (2006). Two *ermB* PCRs were performed as previously described (Böckelmann et al., 2009; Chen et al., 2007). Polymerase chain reactions for β -lactamases were performed. CTX-M was chosen to represent extended-spectrum β -lactamases (Cottell et al., 2013), and CMY-2 was chosen to represent class C β -lactamases (Kozak et al., 2009). Samples that were positive in the qualitative PCR for *tet*(A) and *sul*(I) genes were quantified using QuantiTect SYBR Master Mix (Qiagen) or 5Prime RealMasterMix SYBR ROX (5Prime) with the same primers used for quantitative PCR. All quantitative PCR assays were performed in triplicate. Reported values were determined by comparison with a standard curve for each of the assayed genes and normalized based on grams dry weight of the original soil sample. Primer sequences, PCR recipes, and thermocycling conditions are listed in Supplemental Tables S1 and S2.

Data Analysis

Microbial abundance data (coliforms, enterococci, R2A counts, R2A + tetracycline, and R2A + cefotaxime) were log transformed before statistical analysis. The GLM procedure of SAS (SAS Institute, 2008) was used to determine differences between prairies for the various microbial abundances. For each microbial abundance, differences between specific prairies were determined using Tukey's Studentized range test. Significant correlations ($P < 0.05$) between microbial abundances and various soil and physical parameters were identified using Pearson correlation coefficients. Principal components analysis using the PRINCOMP procedure of SAS was used to determine if the prairies clustered into discrete groups based on the presence of 17 AR genes in soil DNA extracts.

Results

Resistance in Native Nebraska Prairie Soils

Phenotypically, all 100 native Nebraska prairie soil samples collected in this study contained bacteria that were resistant to tetracycline and cefotaxime (defined as bacterial colonies growing on an agar plate containing $16 \mu\text{g mL}^{-1}$ tetracycline or $4 \mu\text{g mL}^{-1}$ cefotaxime). Genotypically, carriage of tetracycline genes in bulk soil DNA extracts was common. Specifically, 98% of soil samples had at least 1 of the 14 tetracycline genes assayed, and tetracycline resistance genes were found in all 20 prairies. Carriage of sulfonamide resistance genes was common, with 91% of the soil samples positive for either *sul*(I) (91%) or *sul*(I) and *sul*(II) (21%). None of the bulk soils was positive for *sul*(II) only. On the prairie level, the *sul*(I) gene was found in all 20 prairies, and the *sul*(II) gene was found in 13. Principal components analysis indicated that the incidence of the 17 specific genes in soil samples yielded no discrete clusters within

prairies (Supplemental Fig. S2). Prairie #15 (LAA) was identified as a possible outlier influenced by a positive *tet(X)* detection and high incidence of *tet(B)* and *tet(S)*. The *erm(B)* gene was not detected in any of the soil samples, with either of the two *erm(B)* PCR protocols used (only PCR controls were positive). The CTX-M and CMY-2 genes were also not observed in any of the samples. No correlations were found between phenotypic or genotypic AR and soil physical or chemical parameters. Results for soil physical and chemical measurements are presented in Supplemental Table S3. Of note, the term “native” refers to the composition and genetics of the plant species in the prairie and does not indicate whether or not the land was historically cultivated. Although many of the prairie remnants we studied are unplowed (i.e., “virgin” prairie), some may have been cultivated since settlement in the mid-1800s.

Tetracycline Resistance Genes

The average number of tetracycline gene types per bulk soil sample ($n = 100$ soil samples total) was 3 (SD, 1.4), with a range of 0 to 7 positive, out of 14 total tetracycline genes examined. The most frequently detected tetracycline genes in Nebraska prairie soil samples were *tet(D)* ($n = 54$ of 100 bulk soil extracts) and *tet(A)* ($n = 52$), with 25 soils positive for both genes (Fig. 1). Other tetracycline resistance genes that were positive in >25% of the soils were *tet(O)* (38%), *tet(L)* (36%), and *tet(B)* (26%). Of the 100 soil samples tested, two were negative for all tetracycline genes tested, and 11 were positive for only 1 of the 14 genes assayed. One soil sample was positive for seven tetracycline resistance genes, and four were positive for six tetracycline resistance genes. Examining the data on the level of prairie (combining results from all five soil samples at each prairie), 18 and 19 prairies were positive for *tet(A)* and *tet(D)*, respectively, and 17 of the 20 prairies were positive for both *tet(A)* and *tet(D)* genes. The least frequently detected tetracycline genes were *tet(Q)* ($n = 0$) and *tet(X)* (two soil samples from two different prairies).

Statistically significant differences were observed between prairies when measuring the number of tetracycline resistance genes present ($P = 0.01$). Of the 20 prairies, all were positive in the quantitative PCR assay for at least two different tetracycline resistance genes.

Quantitative Polymerase Chain Reaction Assays

Two genes, *tet(A)* and *sul(I)*, were chosen for quantification using quantitative PCR. Bulk soil samples that were positive for these genes in the standard PCR assays were quantified. Results are reported in Table 2. The number of copies in positive samples for each of the assayed genes varied both within and between prairie sites [range, 6.98×10^3 – 2.84×10^5 copies g^{-1} dry soil and 9.63×10^2 – 3.56×10^3 copies g^{-1} dry soil for *tet(A)* and *sul(I)*, respectively]. No correlations were observed between gene copy number and isolate phenotypic properties or between gene copy number and soil physical and chemical properties.

Phenotypic Quantification

Standard soil heterotrophic plate counts were performed on all samples and ranged from 3.5×10^4 to 1.8×10^7 CFU per gram dry weight of soil, with the majority of samples ($n = 76$) containing counts in the 10^5 range. Counts of soil bacteria on plates containing antibiotics ranged from 10^2 to 10^6 and from 10^3 to 10^6 for CR2A and TR2A, respectively. Soil heterotroph and cefotaxime-resistant heterotroph data were both negatively correlated with tetracycline resistance phenotype (Table 3). The percentage of isolates that were resistant to cefotaxime and tetracycline varied considerably within each prairie, with the tightest within-prairie counts being separated by 18% points. Despite this within-prairie variability, statistically significant differences were observed between prairies for plate count data (Table 4). For fecal indicators, 99 of the 100 samples were <1 CFU g^{-1} dry weight of soil for *E. coli*. As a reference, soil to which cattle feedlot runoff had been applied commonly had

Gene	Prairie ID																				# Positive Samples ($n = 100$)	# Prairies Positive ($n = 20$)
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20		
<i>tet(A)</i>	3	2	2	2	4	2	4	0	5	2	4	2	3	4	0	3	2	3	1	4	52	18
<i>tet(B)</i>	2	3	0	2	0	2	0	4	0	0	0	0	0	2	5	2	0	2	2	0	26	10
<i>tet(C)</i>	0	0	1	1	1	0	0	0	0	2	1	1	1	2	0	1	0	0	3	0	14	10
<i>tet(D)</i>	3	5	0	5	2	2	1	2	3	4	4	3	1	4	5	4	3	1	1	1	54	19
<i>tet(E)</i>	1	0	0	0	0	1	0	0	3	1	0	0	1	1	2	0	1	2	1	1	15	11
<i>tet(G)</i>	0	0	0	3	0	0	0	0	0	2	1	0	0	0	2	2	2	1	2	0	15	8
<i>tet(K)</i>	0	1	0	3	0	0	0	0	1	2	0	0	0	0	0	0	0	1	0	0	8	5
<i>tet(L)</i>	1	3	2	1	4	2	0	0	1	1	5	2	1	3	0	2	3	0	0	5	36	15
<i>tet(M)</i>	0	1	0	0	0	0	1	1	0	0	0	4	1	0	1	0	4	0	0	1	14	8
<i>tet(O)</i>	0	3	3	0	2	1	2	2	2	1	4	1	2	3	2	4	4	1	0	1	38	17
<i>tet(Q)</i>	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
<i>tet(S)</i>	0	0	0	0	0	2	1	0	0	0	0	0	0	0	4	1	0	0	4	0	12	5
<i>tet(X)</i>	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	1	2	2
<i>tetA(P)</i>	1	0	0	0	0	1	0	0	0	1	0	1	2	3	2	0	2	1	0	2	16	10
<i>sul(I)</i>	4	5	5	5	4	5	5	5	5	4	5	1	4	5	5	5	5	5	5	4	91	20
<i>sul(II)</i>	0	1	2	2	1	2	1	1	0	2	1	0	2	3	1	0	0	0	2	0	21	13

Fig. 1. Tetracycline and sulfonamide resistance genes detected in native Nebraska prairie soils. At each prairie site, samples were taken at five locations. Each result cell indicates the number of samples (of five total) that were positive for the listed gene. Red indicates all five samples at a prairie contained the assayed gene. Yellow and orange indicate one to four samples from the prairie contained the assayed gene. Green indicates none of the samples from that prairie were positive in our assay.

Table 2. Quantitative polymerase chain reaction listed as average copies of the measured gene per gram dry soil. Values are an average of all positive samples from the indicated prairie.

Prairie ID	<i>tet(A)</i>	<i>sul(I)</i>
1	3.23 E+04	1.89E+03
2	2.36 E+04	1.58 E+03
3	1.63 E+05	2.17 E+03
4	2.39 E+04	1.78 E+03
5	8.58 E+04	1.47 E+03
6	3.44 E+04	1.63 E+03
7	1.27 E+05	1.72 E+03
8	ND†	2.64 E+03
9	1.28 E+05	1.83 E+03
10	1.12 E+05	1.71 E+03
11	1.14 E+05	1.61 E+03
12	3.60 E+04	1.22 E+03
13	7.39 E+04	1.81 E+03
14	1.69 E+04	1.84 E+03
15	ND	1.94 E+03
16	2.09 E+04	1.46 E+03
17	1.05 E+05	1.78 E+03
18	5.05 E+04	2.21 E+03
19	1.97 E+05	1.97 E+03
20	1.24 E+05	1.67 E+03

† None detected.

10^3 CFU g^{-1} dry weight *E. coli* (unpublished data). No statistically significant differences were observed between prairies for total coliform counts. However, differences were observed for *Enterococcus* counts.

Disk Diffusion Assays

For each of the 100 bulk soil samples, three isolates were picked off of nonselective plates and subjected to disk diffusion assays. Of these 300 isolates, five were lost during the freezing process, resulting in disk diffusion data for 295 isolates. Resistance, as defined using the CLSI standard methods and clinical breakpoints, was found to all 12 of the drugs tested (Table 5). On an isolate level, the drugs to which the fewest number of isolates displayed resistance were ciprofloxacin and kanamycin, each with 2% of the tested isolates displaying resistance. The drugs to which the largest number of isolates displayed resistance were ceftriaxone (43%) and ampicillin (33%). Of the tested isolates, only 7% displayed resistance to tetracycline using the standard CLSI definitions. All of the tested isolates were picked off of the nonselective R2A plates.

Had isolates been selected off of the R2A with tetracycline, it is possible that a higher proportion would have displayed clinical resistance. There were 72 isolates (24%) that were resistant to a single antibiotic tested and 141 (48%) that were resistant to two or more antibiotics tested. Examining the data on the prairie level, all 20 prairies had at least one isolate resistant to ceftriaxone and at least one isolate resistant to ampicillin. Nineteen of the 20 prairies had at least one isolate resistant to each cefoxitin and nalidixic acid. No statistically significant differences were observed between prairies for the number of phenotypic resistances displayed.

Discussion

Ungrazed native Nebraska prairie soils contain measurable amounts of antibiotic-resistant bacteria and AR genes. These results are consistent with other studies of AR in soils (Walsh and Duffy, 2013; Guardabassi and Agersø, 2006; Schmitt et al., 2006; Demanèche et al., 2008; Davelos et al., 2004) and are to be expected because antibiotic-resistant bacteria and their genes are a normal and natural part of soils (Cytryn, 2013; D'Costa et al., 2011; Finley et al., 2013). Of note is the high prevalence of sulfonamide resistance genes in the present study (91%). The *sul(I)* gene concentrations measured were fairly consistent across prairies, in the range of 10^3 copies per gram dry soil. Watershed studies have found the concentration of sulfonamide resistance genes to be correlated with human or animal antibiotic use, and the *sul* genes have been proposed as markers of anthropogenic AR gene pollution (Pruden et al., 2006). Data from this current prairie study suggest that sulfonamide resistance gene distributions may be different for land-based and water-based systems. However, because different methodologies were used, it is difficult to compare results directly. Furthermore, the microbial community structure of the soil differs considerably from the microbial community structure of fresh water streams and riverbeds. Because the Genes are correlated, to some degree with taxonomy (Durso et al., 2012), it is not surprising that a sample with a very different microbial community structure also has a different distribution of AR genes. The ungrazed native prairie sites examined in this study provided valuable information on background levels and types of AR that can be expected in southeastern Nebraskan soils in the absence of inputs from food animal agriculture.

The working hypothesis for this study was that geographically clustered ungrazed native Nebraska prairie soils would display similar levels and types of AR. More specifically,

Table 3. Correlation coefficients† of bacterial counts with selected phenotypic and genotypic measures of antibiotic resistance.

Bacterial counts	Tetracycline resistance phenotype	Multidrug resistance phenotype	Tetracycline resistance genotype
Total coliform‡	0.06 (0.52)§	−0.003 (0.20)	−0.25** (0.01)
Enterococcus	0.14 (0.15)	−0.15 (0.13)	0.11 (0.30)
Soil heterotrophs	−0.56*** (<0.0001)	−0.05 (0.65)	−0.14 (0.16)
Cefotaxime-resistant soil heterotrophs	−0.33*** (0.0008)	0.03 (0.74)	0.10 (0.33)
Tetracycline-resistant soil heterotrophs	NA	0.15 (0.12)	−0.11 (0.27)

** Significant at the 0.01 probability level.

*** Significant at the 0.001 probability level.

† A correlation coefficient is significant at the 95% level if $|\text{correlation}| > 0.197$ for $n = 100$ (values in bold).

‡ Total coliforms includes some plant-associated bacteria.

§ Values in parentheses represent the $Pr > |r|$.

Table 4. Bacterial counts as affected by prairie site.

Prairie ID	Total coliforms	<i>Enterococcus</i>	R2A	CR2A	TR2A
log CFU g ⁻¹					
1	2.9a†	3.8a	5.1bcd	4.8bc	5.1abcd
2	3.5a	3.8a	5.2abcd	4.8abc	4.9abcd
3	4.0a	3.1ab	5.9a	5.5ab	5.6a
4	3.8a	3.5ab	5.7abc	5.4abc	5.4a
5	4.2a	3.2ab	5.4abcd	5.3abc	5.4a
6	4.0a	3.3ab	5.4abcd	5.1abc	5.0ab
7	3.6a	3.8a	5.7ab	5.6a	5.2ab
8	2.6a	3.1ab	5.0cd	4.7c	5.3ab
9	2.8a	2.8ab	5.0bcd	4.9abc	5.1ab
10	3.4a	2.8ab	5.4abcd	5.2abc	4.9ab
11	3.6a	2.8ab	5.2bcd	4.9abc	5.3a
12	3.1a	3.3ab	5.4abcd	5.0abc	5.3a
13	3.6a	2.9ab	5.4abcd	5.1abc	4.9ab
14	3.4a	3.4ab	5.3abcd	5.2abc	5.3ab
15	2.2a	3.4ab	5.2abcd	4.6c	5.0ab
16	3.4a	3.3ab	5.0d	4.8abc	5.2ab
17	2.3a	2.4b	5.0d	4.7bc	4.9ab
18	3.8a	3.4ab	5.2abcd	5.0abc	5.1ab
19	3.1a	2.6b	5.5abcd	5.2abc	4.1b
20	2.5a	3.2ab	5.2abcd	4.9abc	5.2ab
LS SE‡	0.4	0.2	0.1	0.2	0.2

† Values followed by different letters differ ($p < 0.05$).

‡ Least squares standard error.

“similar” was defined for this study as (i) the trends for which individual soil samples have “more” resistance would be the same regardless of which phenotypic or genotypic measure of resistance was used and (ii) the absolute measurements of each AR parameter would be clustered around a central value, particularly within each prairie site. Although the results of this study provide some limited examples of similarity within and/or between prairie sites, as a whole the data collected for 14 phenotypic and 19 genotypic measures of resistance assayed here do not support this hypothesis. Instead, the measurements reveal heterogeneity in baseline measures of AR between prairies and even among samples collected from the same prairies. Additionally, if “more resistance” is interpreted to mean a higher percentage of samples displaying the measured trait,

then whether or not one prairie had more resistance than another depended on which phenotypic or genotypic assay was performed. For example, only half of the prairie sites ($n = 10$) were positive for *tet(B)*, compared with 19 of 20 positive prairie sites for *tet(D)*. If resistance was defined as presence or absence of the *tet(B)* gene, different conclusions would have been drawn about the prairie sites than if resistance was determined using only *tet(D)*. Similarly, 100% of the soil samples from this study displayed tetracycline resistance, defined as bacterial colonies growing on an agar plate containing 16 $\mu\text{g mL}^{-1}$ of tetracycline, but only 7% of the bacterial isolates from these same soils displayed resistance to tetracycline using the CLSI disk diffusion assay. As with other heterogeneous soil characteristics, phenotypic and genotypic AR measures can vary, even within

Table 5. Disk diffusion assay summary.

Drug	Drug classification	Percent resistant isolates ($n = 295$)	Percent resistant prairies ($n = 20$)
Amoxicillin with clavulanic acid	β -lactam (penicillin) and enzyme inhibitor	16	85
Ampicillin	β -lactam (penicillin)	33	100
Cefoxitin	β -lactam (2GC†)	27	95
Ceftriaxone	β -lactam (3GC‡)	43	100
Chloramphenicol	amphenicol	10	70
Ciprofloxacin	fluoroquinolone	2	30
Gentamicin	aminoglycoside	10	80
Kanamycin	aminoglycoside	2	25
Nalidixic acid	quinolone	21	95
Streptomycin	aminoglycoside	14	85
Tetracycline	tetracycline	7	50
Sulfamethoxazole + trimethoprim	sulfonamide + enzyme inhibitor	12	70

† Second-generation cephalosporin.

‡ Third-generation cephalosporin.

a fairly restricted space and time. The data collected in this study provide information on the range of values that can be expected in unaffected Nebraska soils. The variety observed in the measured parameters supports pooling of samples and the use of multiple measurement parameters (Singer and Williams-Nguyen, 2014) when evaluating AR in manure-affected soils.

Measuring Antibiotic Resistance in Agriculture

Although a common goal is to reduce AR on farms and in feedlots (Finley et al., 2013), there is little available evidence to help decide what a realistic target would be for reduction. Knowing what kinds and amounts of resistance are in native prairie soils will help to determine the kinds and amounts of resistance that can be affected by agricultural best management practices (agBMPs) (Singer and Williams-Nguyen, 2014) in other southeastern Nebraska soils. With over 4570 cattle feeding operations in Nebraska, it is common to apply cattle manure to the soil. Ideally, data on background levels of AR will be gathered for each of the specific AR parameters being assayed in manure-affected soils. Knowing that over 70% of the 300 isolates collected here from minimally affected soils display resistance (as defined by CLSI disk diffusion assay), with almost half (48%) being multidrug resistant (31% if “multidrug resistant” is defined as resistance to three or more drugs) can help to put into perspective the numbers coming out of studies in agricultural production systems. When setting targets for reduction, it is unlikely that agBMPs can reduce resistance below the background levels.

If it is not possible or practical to collect background data from minimally affected sites such as ungrazed prairies or forest soils (Storteboom et al., 2010; Popowska et al., 2012; Shange et al., 2012), the next best option would be to collect baseline data (i.e., from the soils before manure application) (Cook et al., 2014; McLain and Williams, 2014; Marti et al., 2014) so that it is possible to separate out the types of resistance that are most relevant to agricultural production and human health (Durso and Cook, 2014; Wittum, 2012; Phillips et al., 2004). One question that arises from this study is: When examining AR in agricultural soils, does naturally occurring resistance present the same risks to human health as resistance that has been enriched by the administration of drugs to food animals (Martínez et al., 2015)? In one study, prairie soils were found to have more β -lactamase genes than cultivated fields (Demanèche et al., 2008). Data from the current study reveal widespread phenotypic and genotypic resistance. On a practical level, when developing strategies to minimize the transfer of AR from agroecosystems, we argue that it does not matter whether the resistance was naturally occurring, as in this study, or enriched due to human use.

A common assumption regarding the use of antibiotics in agriculture is that reduced administration of antibiotics to food animals will result in a decrease of AR in the animal feces, leading to a decreased risk to human health (Singer et al., 2003; Marshall and Levy, 2011). However, recent data reveal that feces, even from antibiotic-free animals, can carry multiple types of resistance and enrich for AR in the soil (Udikovic-Kolic et al., 2014; Zhou et al., 2009; Kyselková et al., 2013; Schmitt et al., 2006). Thus, studies in unaffected sites that are not receiving manure

inputs provide a unique insight into the background types and distributions of AR in local agricultural soils.

Ecology of Resistance in Nebraskan Soils

When discussing AR in the environment, it is common to equate resistance not only with antibiotic-resistant bacteria but also with the presence or absence of particular AR genes (Zhang et al., 2013; Martínez et al., 2015; Storteboom et al., 2010; Durso et al., 2011). We looked for correlations between phenotypic and genotypic measures of resistance and common soil physical and chemical parameters but did not identify any relationships between abiotic drivers of microbial communities and the selected measures of antibiotic-resistant bacteria or AR genes. Metagenomic studies of bacterial diversity and function in preagricultural prairie soils also found no links between soil variables and bacterial taxonomy (Fierer et al., 2013).

One shortcoming of environmental studies of AR, including this study, is that “resistance” is commonly defined using clinical terms, which may or may not accurately reflect the relationships and dynamics of environmental isolates or soil communities (Walsh and Duffy, 2013). For example, in this study the CLSI standard methods for disk diffusion assays were used, including using the CLSI interpretive criteria for assigning “resistant,” “intermediate,” or “sensitive” status for each data point. In clinical settings, these categories are linked to the ecology of infectious disease, with “resistance” being strongly correlated with clinical treatment failure (Jorgensen and Ferraro, 2009). The use of a term with environmental isolates or communities implies a threat equivalent to resistance in clinical settings. However, resistance is not generally defined for environmental isolates (Berendonk et al., 2015).

Conclusions

Data from ungrazed Nebraska prairie soils indicate an inherent heterogeneity in natural soils for both phenotypic and genotypic measurements of AR. Resistant bacteria and resistance genes are common, even in native prairie soils with no inputs from domesticated food animals. To sort out the contributions of food-animal antibiotic use on resistance at farms and in the environment, it is therefore essential that background and/or baseline levels of AR be considered and acknowledged when evaluating AR in agroecosystems. Data from agriculturally affected sites must be evaluated in the larger context of naturally occurring levels of resistance for individual production systems and soil types.

Reducing AR in agroecosystems, especially food animal production systems, is an important issue. Data from this study, demonstrating AR occurring naturally in the soil and advocating that assessments of resistance on farms and in manure-affected environments include the collection of background and/or baseline data, do not minimize the importance of identifying, evaluating, and adopting agBMPs that reduce AR in agricultural production systems. Instead, these applied data can be used to more accurately assess the impacts of individual agBMPs on specific measures of AR, allowing us to focus limited resources where they will have the most impact.

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Assessment of Selected Antibiotic Resistances in Ungrazed Native Nebraska Prairie Soils
Supplementary Figure 1. Prairie sites. Shaded counties indicate areas from which samples were collected. Counties, listed from north to south, are Lancaster, Otoe, Johnson, Pawnee, and Richardson.



Assessment of Selected Antibiotic Resistances in Ungrazed Native Nebraska Prairie Soils

Table S1: PCR Primers

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Table S1. Primers used in study.

Primer Name	Sequence	Length (bp)	TM (°C)	Product size (bp)
Ng <i>tet</i> (A) F†	5'-GCT ACA TCC TGC TTG CCT TC -3'	20	63.8	210
Ng <i>tet</i> (A) R†	5'-CAT AGA TCG CCG TGA AGA GG -3'	20	64.6	
Ng <i>tet</i> (B) F†	5'-TTG GTT AGG GGC AAG TTT TG -3'	20	63.6	659
Ng <i>tet</i> (B) R†	5'-GTA ATG GGC CAA TAA CAC CG -3'	20	63.7	
Ng <i>tet</i> (C) F†	5'-CTT GAG AGC CTT CAA CCC AG -3'	20	63.8	418
Ng <i>tet</i> (C) R†	5'-ATG GTC GTC ATC TAC CTG CC -3'	20	63.9	
Ng <i>tet</i> (D) F†	5'-AAA CCA TTA CGG CAT TCT GC -3'	20	63.6	787
Ng <i>tet</i> (D) R†	5'-GAC CGG ATA CAC CAT CCA TC -3'	20	64.0	
Ng <i>tet</i> (E) F†	5'-AAA CCA CAT CCT CCA TAC GC -3'	20	63.6	278
Ng <i>tet</i> (E) R†	5'-AAA TAG GCC ACA ACC GTC AG -3'	20	63.7	
Ng <i>tet</i> (G) F†	5'-CAG CTT TCG GAT TCT TAC GG -3'	20	63.5	844
Ng <i>tet</i> (G) R†	5'-GAT TGG TGA GGC TCG TTA GC -3'	20	63.6	
Ng <i>tet</i> (K) F†	5'-TCG ATA GGA ACA GCA GTA -3'	18	54.0	169
Ng <i>tet</i> (K) R†	5'-CAG CAG ATC CTA CTC CTT -3'	18	54.1	
Ng <i>tet</i> (L) F†	5'-TCG TTA GCG TGC TGT CAT TC -3'	20	63.9	267
Ng <i>tet</i> (L) R†	5'-GTA TCC CAC CAA TGT AGC CG -3'	20	64.0	
Ng <i>tet</i> (M) F†	5'-GTG GAC AAA GGT ACA ACG AG -3'	20	59.6	406
Ng <i>tet</i> (M) R†	5'-CGG TAA AGT TCG TCA CAC AC -3'	20	60.6	
Ng <i>tet</i> (O) F†	5'-AAC TTA GGC ATT CTG GCT CAC -3'	21	62.1	515
Ng <i>tet</i> (O) R†	5'-TCC CAC TGT TCC ATA TCG TCA -3'	21	65.2	
Ng <i>tet</i> (Q) F†	5'-TTA TAC TTC CTC CGG CAT CG -3'	20	63.7	904
Ng <i>tet</i> (Q) R†	5'-ATC GGT TCG AGA ATG TCC AC -3'	20	63.9	
Ng <i>tet</i> (S) F†	5'-CAT AGA CAA GCC GTT GAC C -3'	19	60.7	667
Ng <i>tet</i> (S) R†	5'-ATG TTT TTG GAA CGC CAG AG -3'	20	63.8	
Ng <i>tet</i> (X) F†	5'-CAA TAA TTG GTG GTG GAC CC -3'	20	63.7	468
Ng <i>tet</i> (X) R†	5'-TTC TTA CCT TGG ACA TCC CG -3'	20	63.7	
Ng <i>tetA</i> (P) F†	5'-CTT GGA TTG CGG AAG AAG AG -3'	20	63.7	676
Ng <i>tetA</i> (P) R†	5'-ATA TGC CCA TTT AAC CAC GC -3'	20	63.3	
Pei <i>sul</i> (I) F‡	5'-CGC ACC GGA AAC ATC GCT GCA C-3'	22	63.8	163
Pei <i>sul</i> (I) R‡	5'-TGA AGT TCC GCC GCA AGG CTC G-3'	22	64.6	
Pei <i>sul</i> (II) F‡	5'-TCC GGT GGA GGC CGG TAT CTG G-3'	22	65.1	191
Pei <i>sul</i> (II) R‡	5'-CGG GAA TGC CAT CTG CCT TGA G-3'	22	60.6	
Böckelmann <i>erm</i> (B) F§	5'- GCA TTT AAC GAC GAA ACT GGC T -3'	22	56.2	573
Böckelmann <i>erm</i> (B) R§	5'-GAC AAT ACT TGC TCA TAA GTA ATG GT-3'	26	53.2	
Chen <i>erm</i> (B) F¶	5'-GAT ACC GTT TAC GAA ATT GG-3'	20	49.0	364
Chen <i>erm</i> (B) R¶	5'-GAA TCG AGA CTT GAG TGT GC-3'	20	53.4	
Kozak CMY-2 F#	5'-GAC AGC CTC TTT CTC CAC A-3'	19	54.3	1015
Kozak CMY-2 R#	5'-TGG AAC GAA GGC TAC GTA-3'	18	52.9	
Cottell CTX-M F††	5'-CCG CTG CCG GTY TTA TC-3'	17	55.1	490-520
Cottell CTX-M R††	5'-ATG TGC AGY ACC AGT AA-3'	17	49.3	

- † Ng, L.K., I. Martin, M. Alfa, and M. Mulvey. 2001. Multiplex PCR for the detection of tetracycline resistant genes. *Mol. Cell. Probes*. 15:209-215.
- ‡ Pei, R., K. Sung-Chul, K.H. Carlson, and A. Pruden. 2006. Effect of River Landscape on the sediment concentrations of antibiotics and corresponding antibiotic resistance genes (ARG). *Water Res.* 40:2427-2435.
- § Böckelmann, U., H. Dörries, M.N. Ayuso-Gabella, M. Salgot de Marçay, V. Tandori, C. Levantesi, C. Mascioppinto, E. Van Houtte, U. Szewzyk, T. Wintgens, and E. Grohmann. 2009. Quantitative PCR Monitoring of Antibiotic Resistance Genes and Bacterial Pathogens in Three European Artificial Groundwater Recharge Systems. *Appl. Environ. Microb.* 75:154-163.
- ¶ Chen, J., Z. Yu, F.C. Michel, Jr., T. Wittum, and M. Morrison. 2007. Development and application of real-time PCR assays for quantification of *erm* genes conferring resistance to macrolides- lincosamides-streptogramin B in livestock manure and manure management systems. *Appl. Environ. Microb.* 73:4407-4416.
- # Kozak, G.K., P. Boerlin, N. Janecko, R.J. Reid-Smith, and C. Jardine. 2009. Antimicrobial resistance in *Escherichia coli* isolates from swine and wild small mammals in the proximity of swine farms and in Natural Environments in Ontario, Canada. *Appl Environ Microbiol.* 75:559-566.
- †† Cottell, J.L., N. Kanwar, L.Castillo-Courtade, G. Chalmers, H.M. Scott, B. Norby, G.H. Loneragan, and P. Boerlin. 2013. Blactx-M-32 on and *incN* plasmid in *Escherichia coli* from beef cattle in the United States. *Appl Environ Microbiol.* 79:1096-1097.

Assessment of Selected Antibiotic Resistances in Ungrazed Native Nebraska Prairie Soils

Table S2: PCR Recipes and Thermocycling Conditions

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Qualitative PCR

Resistance	Genes	PCR Recipe Used	Thermocycling Conditions Used	Reference
Tetracycline	A, B, C, D, E, G, K, L, M, O, Q, S, X, A(P)	12.5 µl of Jumpstart Red Taq was combined with 0.5 µl of each primer (100mM), 1 µl of extracted DNA, and reagent grade water to bring the total reaction volume to 25 µl.	1 cycle of 94°C for 5 min; 35 cycles of 94°C for 1 min, 55°C for 1 min and 72°C for 90 seconds; and one cycle of 72°C for 5 minutes.	Ng et al. 2001
Sulfonamide	sul(I)	9 µl of 5PrimeMasterMix (5Prime, Gaithersburg MD), 0.4 µl of each primer (100mM), 1 µl of template DNA, and reagent-grade water in a final reaction volume of 20 µl.	PCR were 1 cycle of 95°C for for 15 min; 50 cycles of 95 for 15 sec, 65°C for 30 sec, and 72°C for 30 sec.	Pei et al. 2006
	sul (II)	9 µl of 5PrimeMasterMix (5Prime, Gaithersburg MD), 0.4 µl of each primer (100mM), 1 µl of template DNA, and reagent-grade water in a final reaction volume of 20 µl.	PCR were 1 cycle of 95°C for for 15 min; 50 cycles of 95 for 15 sec, 57.5°C for 30 sec, and 72°C for 30 sec.	Pei et al. 2006
β-lactamases	CTX-M, CMY-2	10ul of Jumpstart Red Taq Master Mix (Sigma, St. Louis MO) was combined with 0.2ul CMY2 primers (200mM) and 0.3ul CTX-M primers (200mM), 1ul of extracted DNA and reagent-grade water for a total reaction volume of 20ul.	One cycle consisting of 15 min at 94°C, 30 cycles consisting of 1 min at 94°C, 1 min at 55°C, and 1 min at 72°C, and one cycle consisting of 10 min at 72°C.	Cottell et al., 2013; Kozak et al., 2009

Quantitative PCR

Resistance	Genes	PCR Recipe Used	Thermocycling Conditions Used	Reference
Tetracycline	tet (A)	QuantiTect SYBR Master Mix (Qiagen, Valencia CA)	1 cycle of 94°C for 15 min; 35 cycles of 94°C for 1 min, 55 for 1 min, 72°C for 90 sec; 1 cycle of 72°C for 5 min	Ng et al. 2001
Sulfonamide	sul (I)	5Prime RealMasterMix SYBR ROX (5Prime, Gaithersburg MD)	95°C for 15 min; 50 cycles of 95°C for 15 sec, 65°C for 30 sec, 72°C for 30 sec	Pei et al., 2006

Cottell, J.L., N. Kanwar, L.Castillo-Courtade, G. Chalmers, H.M. Scott, B. Norby, G.H. Loneragan, and P. Boerlin. 2013. Blactx-M-32 on and incN plasmid in *Escherichia coli* from beef cattle in the United States. *Appl Environ Microbiol.* 57:10-96-1097.

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Ng, L.K., I. Martin, M. Alfa, and M. Mulvey. 2001. Multiplex PCR for the detection of tetracycline resistant genes. *Mol. Cell. Probes.* 15:209-215.

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Assessment of Selected Antibiotic Resistances in Ungrazed Native Nebraska Prairie Soils

Table S3: Soil Physical and Chemical Data

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Prairie ID	Moisture Content	pH	Organic Matter		Potassium K (ppm)	Sulfate-S S (ppm)	Zinc Zn (ppm)	Iron Fe (ppm)	Manganese Mn (ppm)	Copper		Magnesium Mg (ppm)	Sodium KCl NO3-		Mehlich P-III (ppm)
			(LOI %)							Cu (ppm)	Ca (ppm)		Na (ppm)	N NO3 (ppm)	
1	9.74	6.98	4.38	253	8.4	1.4	62.8	11.4	1.018	2198.2	370.8	11.2	1.34	11.88	5.6
2	12.63	6.92	3.94	210.6	8.2	0.932	64.88	8.68	1.15	1917.6	474	24.6	1	11.22	3
3	24.33	6.76	5.82	211.2	14.4	1.424	77.48	22.86	1.546	3072.8	347	33	1.86	12.98	4.2
4	20.33	6.76	3.52	254	17.6	0.93	91.9	20.46	2.046	3575.6	640.6	19	2.88	14	13.2
5	21.84	6.74	5.28	165.2	11	1.144	83.32	14.28	1.494	1993.6	347.4	13.2	1.02	12.38	2.4
6	20.94	6.88	6.2	210	8	1.1	65.12	23.36	1.202	3687.4	394.6	54	2.04	10.94	2.4
7	17.87	6.92	6.76	247.8	8.4	2.268	57.02	15.28	1	2758.2	290.2	17.2	2.9	9.96	3.8
8	10.62	6.87	5.1	94.8	8.4	1.524	108.7	11.78	1.412	2210.2	407	31.2	2.44	12.24	2.4
9	12.94	6.97	5.14	220.2	7.4	1.212	60.7	12.76	1.046	2125.6	303.2	5	1.32	9.36	4.2
10	14.44	6.78	5.6	162.4	10	1.22	44.74	15.82	1.222	3597	228.6	13.2	4.46	12.46	3.4
11	16.32	6.81	4.72	114	7.2	1.292	81.58	10.56	1.28	2094.6	278.8	15.6	1.14	11.5	2.4
12	12.63	6.88	4.82	246.8	9.2	1.394	73.5	8.3	0.984	1763.6	339.2	9.6	1.36	13.42	4.6
13	20.10	6.92	4.84	128	7.2	1.23	77.2	16.5	1.354	2914	246.4	18.6	1.06	11.68	2.8
14	23.38	6.70	6.06	207.4	13.4	0.912	40.06	30.1	1.412	3715.2	271	21	1.28	14.74	2.8
15	10.10	6.98	5.28	313.8	10.6	1.042	39.02	11.3	1.108	2949.8	397	8	1.26	13.76	4.6
16	11.53	6.88	4.64	186.8	8.2	1.512	164.9	9.52	1.226	1850.6	384.6	32.6	1.1	10.18	3.5
17	11.51	6.90	5.22	161.6	10.6	1.542	97.18	13.98	1.27	2301.2	417.4	18.4	1.72	12.16	3.8
18	15.12	6.87	5.08	283	7.8	0.976	67.5	13.36	1.228	2491.2	508.4	9	1.46	16.98	4.4
19	24.34	7.04	5.56	219.4	11.4	0.564	29.44	23.26	0.87	5317.8	101.4	5.8	2.14	14.66	3.4
20	11.07	6.89	4.38	141.8	9.2	1.544	84.2	12.26	1.16	1778.8	371.4	42	2.18	12.26	3.4

ND = Not detected