

Good morning, I would like to thank IAFP for inviting me to present our work today.

Outline

- What is One Health
- Surveillance of multiple components of antimicrobial resistance (AR) in a watershed
 - Isolate bacteria and assay for AR and AR genes
 - qPCR of AR genes and source tracking markers
 - Quantitation of 26 antibiotics
- Results
- Conclusions
- How does this information effect the One Health approach to antimicrobial resistance?



Today, I am going to talk about the research completed by Gabi Cho during her PhD and postdoc years, which is on the bacteria found in the environment, particularly surface water.

So, why am I interested in water? Why am I interested in the environment? One Health approach states that the health of humans and animals are connected to the health of environment, and we have to work together in order to understand all three compartments and thereby achieving optimal health outcomes for all. This One health approach is also used to tackle the global problem of AR. However, the environment was a big data gap compared to human and animals, and we decided to study bacteria present in the water environment, especially pathogenic bacteria and AR bacteria, to be able to see the whole picture of what is going on.



Sohyun Cho, or Gabi was a new graduate student and was interested in studying antimicrobial resistant bacteria in water because when her family were missionaries in India, she saw her friends and family struggle to get safe water and often got sick from the water.

I said we don't do water, but I remember we had done water with Rick Meinersmann and that the our local watershed had high levels of bacteria and AR.

We showed this to our lead scientist, Charlene Jackson, and she said the environment was a big data gap, so we presented this to our NPL James Lindsey, and he liked the idea and approved the research because the one health approach to AR required data from the environment.

So we developed a collaboration with Erin Lipp and Elizabeth Ottesen at UGA and with the upper Oconee watershed network volunteers.



This map shows that Salmonella infection is highest in Southeastern US as depicted by the dark green color.

In 2013, Georgia had the highest case rate of Salmonella infections with 26.05 cases per 100,000 compared to the US average of 14.5.



This graph shows a seasonal trend in Salmonella infections. Y-axis represents the number of Salmonella infections reported to CDC and X-axis represents the month the specimens were collected. There is a clear pattern repeating every year with more infections during summer months.



This graph shows a seasonal trend in Salmonella infections.

Y-axis represents the number of Salmonella infections reported to CDC and X-axis represents the month the specimens were collected.

There is a clear pattern repeating every year with more infections during summer months.

Why would a food borne disease be seasonal?

There simply can't be that much potato salad, plus, most cases reported are not associated with an outbreak and are what we call sporadic cases with no known source.

So what are some other reasons summer months it goes from less than 400 to nearly three times that?

Other things people do in the warm months is outdoor recreation, and a lot of that has contact with surface water, including swimming, fishing, boating, etc.

So, why not look for the cause of this increase in the environment?



We not only isolated ARB from surface water to characterize the bacteria that are culturable and express their ARGs, but also quantified the total ARGs present within the whole bacterial populations in the watershed, including those that have not been expressed.

In addition, antibiotics that are important to human and veterinary medicine were measured to investigate the occurrence and distribution of these antibiotics in aquatic environments.

As WWTPs have been proposed to be the hotspots for the emergence of ARB and from where these bacteria are spread into the natural environment, influents and effluents from three WWTPs located within the watershed and whose effluents flow into the streams within the watershed were included in the analyses.



As you all know, Salmonella are pathogenic bacteria and while E. coli and Enterococcus are commensal bacteria, certain E. coli, such as O157:H7, and E. faecalis and E. faecium, are pathogenic and can infect humans and animals. Also, these bacteria were chosen by NARMS or National Antimicrobial Resistance Monitoring System as sentinel organisms for monitoring AR in food animals, retail meats, and humans, so we chose to study these bacteria to monitor AR in surface water as well.

Surface water receives contamination from the surroundings and human and animals can be exposed to contaminants in the water through recreational activities, drinking, and consuming fruits and vegetables irrigated with contaminated water. So the question I had was- Is surface water a reservoir of pathogenic and AR bacteria that can be transmitted to or from humans and animals?



Read slide



This map shows the entire Upper Oconee Watershed, with land use shown in these colors, and the sampling sites shown as black and white circles.

The waterways we sample merge to form the Oconee River which then joins the Altamaha River which flows into the Atlantic Ocean south of Sapelo Sound.



UWON volunteers. Ahn Nguyen. Gabi. 1 liter bottle.



We got about 30 to 100 water samples for each water collection four times a year, depending on the number of volunteers to help us collect water samples. Some of these sites were adjacent to animal farms, wastewater treatment plants, and residential areas with septic tanks.

We filtered the water samples and isolated Salmonella, E. coli and Enterococcus using selective media.



We not only isolated ARB from surface water to characterize the bacteria that are culturable and express their ARGs, but also quantified the total ARGs present within the whole bacterial populations in the watershed, including those that have not been expressed.

In addition, antibiotics that are important to human and veterinary medicine were measured to investigate the occurrence and distribution of these antibiotics in aquatic environments.

As WWTPs have been proposed to be the hotspots for the emergence of ARB and from where these bacteria are spread into the natural environment, influents and effluents from three WWTPs located within the watershed and whose effluents flow into the streams within the watershed were included in the analyses.

	number of	number of % of positive sites (no. of isolates recove			
sampling season	samples	Salmonella	E. coli	Enterococcu	
Winter 2015	30	70.0 (59)*	96.7 (56)*	96.7 (58)*	
Spring 2015	100	68.0 (153)*	99.0 (99)	93.0 (93)	
Summer 2015	33	81.8 (66)*	97.0 (46)*	100.0 (196)*	
Fall 2015	59	30.5 (37)*	100.0 (59)	98.3 (58)	
Winter 2	I			(41)	
Spring 2	1	052 water s	amnles	(87)	
Summer 2				(27)	
Fall 201 7	0 1% nosi [.]	tive for Salm	o <i>nella</i> (n = 1.)	796) (77)	
	0.1/0 0001				
Winter 2	00 E% n/	ositivo for E	coli (n - 1.10)	(11)	
Winter 2 Spring 2	99.5% p	ositive for E.	<i>coli</i> (n = 1,10	3) (51) (94)	
Winter 2 Spring 2 Summer 2 98	99.5% positi 99.5% positi	ositive for <i>E.</i>	<i>coli</i> (n = 1,103 <i>coccus</i> (n = 1	3) (51) (94) (40)	
Winter 2 Spring 2 Summer 2 Fall 201	99.5% positi 99.5% positi	ositive for <i>E.</i> ive for <i>Enterc</i>	<i>coli</i> (n = 1,103 <i>coccus</i> (n = 1	3) (51) (94) (40) (43) (43)	
Winter 2 Spring 20 Summer 2 Fall 201 Winter 2018	99.5% po 8.9% positi	ositive for <i>E.</i> ive for <i>Enterc</i> 48.8 (57)*	coli (n = 1,103) coccus (n = 1 $^{100}(41)$	3) (51) (94) (40) (43) 97.6 (40)	
Winter 2 Spring 20 Summer 2 Fall 201 Winter 2018 Spring 2018	99.5% po 8.9% positi	48.8 (57)* 59.5 (74)*	$\frac{100 (41)}{100 (42)}$	3) (51) (94) (40) (43) 97.6 (40) 100 (42)	
Winter 2 Spring 20 Summer 2 Fall 201 Winter 2018 Spring 2018 Summer 2018	99.5% po 8.9% positi 41 42 44	48.8 (57)* 59.5 (74)* 93.2 (94)*	$\frac{100 \ (41)}{100 \ (42)}$	3) (51) (94) (40) (43) 97.6 (40) 100 (42) 100 (44)	
Winter 2 Spring 2 Summer 2 Fall 201 Winter 2018 Spring 2018 Summer 2018 Fall 2018	99.5% positi 8.9% positi 41 42 44 44 44	48.8 (57)* 59.5 (74)* 93.2 (94)* 61.4 (55)*	$\frac{100 \ (n = 1, 103)}{100 \ (41)}$ $\frac{100 \ (41)}{100 \ (42)}$ $\frac{100 \ (44)}{97.7 \ (43)}$	3) (51) (94) (94) (40) (43) 97.6 (40) 100 (42) 100 (44) 100 (44)	
Winter 2 Spring 20 Summer 2 98 Fall 201 Winter 2018 Summer 2018 Fall 2018 Winter 2018	99.5% positi 41 42 44 44 44 44	48.8 (57)* 59.5 (74)* 93.2 (94)* 61.4 (55)* 59.1 (71)*	$\begin{array}{c} \text{coli} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{coccus} (n =$	3) 50) 51) 51) 94) 40) 40) 43) 97.6 (40) 100 (42) 100 (44) 100 (44) 100 (44)	
Winter 2 Spring 20 Summer 2 Fall 201 Winter 2018 Spring 2018 Summer 2018 Fall 2018 Winter 2019 Spring 2019	99.5% positi 41 42 44 44 44 44 44 41	48.8 (57)* 59.5 (74)* 93.2 (94)* 61.4 (55)* 59.1 (71)* 78.0 (99)*	$\begin{array}{c} \text{coli} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{coccus} (n =$	3) 50) 51) 94) 51) 94) 97.6 40) 43) 97.6 40) 43) 97.6 40) 43) 100 42) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 44) 100 45) 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100 100	
Winter 2 Spring 20 Summer 2 Fall 201 Winter 2018 Spring 2018 Summer 2018 Fall 2018 Winter 2019 Spring 2019 Summer 2019	99.5% positi 41 42 44 44 44 44 41 19	48.8 (57)* 59.5 (74)* 93.2 (94)* 61.4 (55)* 59.1 (71)* 78.0 (99)* 89.5 (40)*	$\begin{array}{c} \text{coli} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{coccus} (n $	3) (51) (40) (43) 97.6 (40) 100 (42) 100 (44) 100 (44) 100 (44) 100 (44) 100 (41) 100 (41) 100 (19)	
Winter 2 Spring 2 Summer 2 Fall 201 Winter 2018 Spring 2018 Summer 2018 Fall 2018 Winter 2019 Spring 2019 Summer 2019 Fall 2019	99.5% positi 8.9% positi 41 42 44 44 44 44 41 19 45	48.8 (57)* 59.5 (74)* 93.2 (94)* 61.4 (55)* 59.1 (71)* 78.0 (99)* 89.5 (40)* 60.0 (64)*	$\begin{array}{c} \text{coli} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{cocccus} (n = 1, 103) \\ \text{coccus} (n $	3) (51) (94) (40) (43) 97.6 (40) 100 (42) 100 (44) 100 (44) 100 (44) 100 (44) 100 (41) 100 (19) 100 (5)	

Now let's go over the results. This is the result of the 21 water collections we have had until we had to stop due to COVID pandemic.

A total of 1,052 water samples were collected, and almost all of the water samples were positive for E. coli and Enterococcus, and about 70% of all the water samples were positive for Salmonella. More than a thousand isolates were recovered for each of the bacteria, but due to limited time we have today, I am going to talk about just a few selected isolates that might be of interest.



For E. coli, I took my 750 E. coli isolates from the first 3 years of my study and determined their phylogenetic groups, pathogenic types, and AR using the methods mentioned earlier. After that, I selected 34 AR E. coli isolates to run PFGE, AR gene PCR, plasmid replicon typing, integron PCR, MLST, and WGS for 6 selected isolates.

And these are the papers that came out of these data:

-Prevalence and characterization of *Escherichia coli* isolated from the Upper Oconee Watershed in Northeast Georgia

-Genetic Characterization of Antimicrobial-Resistant *Escherichia coli* Isolated from a Mixed-Use Watershed in Northeast Georgia, USA

-Genomic Analysis of Multidrug-Resistant *Escherichia coli* from Surface Water in Northeast Georgia, United States: Presence of an ST131 Epidemic Strain Containing *bla*CTX-M-15 on a Phage-Like Plasmid



Now changing to Enterococcus; I took 865 isolates from the first 3 years of my study and determined their species and AR phenotypes. And then, I have selected 51 MDR Enterococcus isolates from the first 2 years that are resistant to 3 or more antimicrobial drugs to run AR gene PCR and plasmid replicon typing.

Also, 32 isolates that were resistant to daptomycin and tigecycline, which are fairly new drugs, were selected for WGS.

And these are the papers that came out of these data:

-Diversity and antimicrobial resistance of *Enterococcus* from the Upper Oconee Watershed, Georgia

-Antimicrobial Resistance Gene Detection and Plasmid Typing Among Multidrug Resistant Enterococci Isolated from Freshwater Environment



Next is Salmonella. I took my 1190 isolates from the first 3 years of the study and determined their serotypes, PFGE patterns, and AR. And then, I selected 52 AR Salmonella isolates to run AR gene PCR, plasmid replicon typing, integron PCR, and WGS for 4 MDR isolates.

And this is the paper that came out of these data: Analysis of *Salmonella enterica* Isolated from a Mixed-Use Watershed in Georgia, USA: Antimicrobial Resistance, Serotype Diversity, and Genetic Relatedness to Human Isolates.

	Sa	Imonelle	a se	rotype	
Agbeni	1	Havana	3	Newport	9
Agona	1	I 4,[5],12:b:-	53	Oranienburg	3
Anatum	36	I 4,[5],12:i:-	3	Orion	4
Aqua	7	Infantis	30	Ouakam	1
Baildon	2	Inverness	5	Paratyphi_B_varL-tartrate+	1
Bareilly	29	Javiana	3	Rubislaw	15
Berta	3	Kentucky	5	Saintpaul	4
Braenderup	39	Kiambu	3	Schwarzengrund	9
Brandenburg	1	Litchfield	4	Senftenberg	4
Brazil	1	Liverpool	2	Soerenga	:
Cerro	2	Luciana	2	Tennessee	1
Cubana	10	Mbandaka	9	Thompson	4
Derby	1	Meleagidis	2	Typhimurium	1
Enteritidis	1	Mississippi	7	Worthington	
Gaminara	25	Montevideo	113	Untypable	
Give/ Give var.	57	Muenchen	270	Salmonella arizonae/ diarizonae	1
Hartford	87	Muenster	16	Salmonella houtenae	2
	Total			1190	

These are all the Salmonella serotypes we obtained from surface water with the number of isolates for each serotype, and the 5 most common serotypes we see are Hartford, Montevideo, Muenchen, Newport, and Rubislaw. You can see that some of the serotypes found in clinical isolates are also found in water, including Enteritidis, Infantis, Typhimurium, and Newport.



These are the PFGE patterns of the 2 most common serotypes, Muenchen and Rubislaw. This shows that each Salmonella serotype presents a high degree of strain diversity.

I compared my environmental isolates with human isolates on CDC PulseNet database and about half of my isolates had indistinguishable PFGE patterns as human clinical isolates, which means they could be clones. There were several incidences where the same *Salmonella* strains with the same PFGE patterns were simultaneously recovered from both surface water and humans in the surrounding area, suggesting a potential epidemiologic association between the aquatic environment and human infections.

Strain diversity of *Salmonella enterica* subspecies enterica serotypes

Serotype	total no. of isolates	% of total isolates	no. of samplings recovered	no. of PFGE patterns
Muenchen	270	22.7%	11	141
Rubislaw	153	12.9%	12	97
Montevideo	113	9.5%	11	31
Newport	92	7.7%	10	31
Hartford	87	7.3%	11	20
Give	57	4.8%	8	31
I 4,[5],12:b:-	53	4.5%	9	17
Braenderup	39	3.3%	7	11
Anatum	36	3.0%	5	7
Infantis	34	2.9%	8	13
Oranienburg	32	2.7%	3	3
Bareilly	29	2.4%	8	14
Gaminara	25	2.1%	7	18
Typhimurium	19	1.6%	5	14
Muenster	16	1.3%	5	6
Cubana	10	0.8%	3	6

A lot of PFGE patterns!

AR Salmonella (n= 52) recovery	rate 4.4%		
AR profiles ^a	No. of resistances	No. of isolates	Serotypes
Pan-susceptible	0	1138	
Nal	1	3	Muenster [3]
Sul	1	1	Montevideo [1]
Tet	1	5	Muenster [3], Muenchen [1], Gaminara [1]
StrTet	2	1	Kentucky [1]
SulTet	2	2	27 S. Oranienburg
StrSulTet	3	6	- 2017 Fall
SulTetTri	3	5	- 10 different sites
AmpChlSulTetTri	5	1	- 100% identical PFGE patterns
AmoAmpFoxTioAxoChlGenStrSulTet	10	27	Oranienburg [27]
AmoAmpFoxTioAxoChlStrSulTetTri	10	1	Newport [1]

This is the result of the susceptibility testing of Salmonella isolates. The recovery rate of AR Salmonella was 4.4% with 52 AR isolates. We have 1 S. Newport, which is resistant to 10 different drugs, and 27 S. Oranienburg, which are also resistant to 10 drugs. Interestingly, these S. Oranienburg isolates were all isolated from the 2017 Fall collection from 10 different sites, but they all seem to be clones with the 100% identical PFGE patterns.



The 52 Salmonella isolates were selected for further testing and this is the dendrogram of the R Salmonella with PFGE patterns, AR phenotypes, replicon types, and serotypes. We can see some clones, including 27 S. Oranienburg that were all isolated in the same season but from different sites. This is the MDR S. Newport with resistance to 10 different drugs and it has an A/C plasmid which is a large plasmid usually associated with MDR. MDR Salmonella Newport with A/C plasmid has caused several outbreaks in humans and cattle, so these bacteria were expected to be isolated from humans and animals, but not from surface water, but in fact, this isolate had a matching PFGE pattern as an outbreak strain.

Isolate (serotype)	Gene ^a	location	Isolate (serotype)	Gene ^a	location
78 TX	(AGly) aadA2	integron (A/C)	561 TX	(AGly) aadA2	HI2
(Newport)	(Bla) bla _{CMY-2}	A/C	(Derby)	(Bla) bla _{TEM-1}	HI2
	(Tmt) dfrA12	integron (A/C)		(Tmt) dfrA12	HI2
	(Phe) floR	A/C		(Phe) floR	HI2
	(AGly) strA	A/C		(Sul) sull	HI2
	(AGly) strB	A/C		(Tet) tetB	HI2
	(Sul) sul1	integron (A/C)	647 GB	(AGly) aadB	A/C
	(Sul) sul2	A/C	(Oranienberg)	(AGly) aph3-Ia	A/C
	(Tet) tetA	A/C		(Bla) bla _{CMY-2}	A/C
	(Tet) tetR	A/C		(Bla) bla _{TEM-1}	A/C
256 GB	(Tmt) dfrA15	Ν		(Phe) cmlA5	A/C
(Montevideo)	(Sul) sull	Ν		(AGly) strA	A/C
	(Tet) tetA	Ν		(AGly) strB	A/C
	(Tet) tetR	Ν		(Sul) sul2	A/C
				(Tet) tetA	A/C
				(Tet) tetR	A/C

4 isolates were selected for WGS and the locations of their resistance genes were identified. You can see that all of the isolates had their AR genes on specific plasmids while 3 AR genes of the MDR S. Newport isolate were located on an integron within the plasmid.

I was able to sequence only a set of 4 Salmonella isolates due to limited resources, but FDA is going to sequence all our 1,800 Salmonella isolates for the GenomeTrackr database which is like the CDC PulseNet database but instead of clinical isolates, it is a database of the WGS of the non-clinical isolates and managed by FDA. So we may have more interesting and comprehensive data in near future.

Serotype	total no. of isolates	no. of isolates with matching PFGE patterns (%)	Serotype	total no. of isolates	no. of isolates wit matching PFGE patterns (%)
Agona	1	1 (100%)	Liverpool	2	2 (100%)
Anatum	36	8 (22.2%)	Luciana	2	0
Aqua	7	0	Mbandaka	9	9 (100%)
Baildon	2	2 (100%)	Meleagidis	2	2 (100%)
Bareilly	30	18 (60%)	Mississippi	7	6 (85.7%)
Berta	3	1 (33.3%)	Montevideo	113	90 (79.6%)
Brandenburg	1	0	Muenchen	270	34 (12.6%)
Braenderup	39	31 (79.5%)	Muenster	17	13 (76.5%)
Brazil	1	0	Newport	99	77 (77.8%)
Cerro	2	2 (100%)	Oranienburg	32	32 (100%)
Cubana	14	1 (7.1%)	Orion	4	0
Derby	1	1 (100%)	Ouakam	1	0
Enteritidis	1	1 (100%)	Paratyphi_B_varL-tartrate+	• 1	0
Gaminara	25	0	Rough O:i:-	1	0
Give	57	2 (3.5%)	Rubislaw	153	28 (18.3%)
Hartford	86	75 (87.2%)	Saintpaul	4	4 (100%)
Havana	3	3 (100%)	Schwarzengrund	9	5 (55.6%)
Infantis	22	15 (68.2%)	Senftenberg	4	4 (100%)
Inverness	5	2 (40%)	Soerenga	5	5 (100%)
I 4,[5],12:b:-	53	40 (75.5%)	Tennessee	2	2 (100%)
I 4,[5],12:i:-	3	3 (100%)	Thompson	4	4 (100%)
Javiana	3	3 (100%)	Typhimurium	19	16 (84.2%)
Kentucky	5	1 (20%)	Worthington	1	Ò Ó
Kiambu	3	0	subspecies III (III_48:g,z51:-)	19	1 (5.3%)
Kintambo	1	1 (100%)	subspecies IV	2	Ì0 Ú

Table showing salmonella serotype (version 3) and PFGE patterns having indistinguishable PFGE patterns as clinical isolates.

About half (46.1%) of the isolates had PFGE patterns indistinguishable from human clinical isolates in the CDC PulseNet database.

There were several incidences where the same *Salmonella* strains with the same PFGE patterns were simultaneously recovered from both surface water and humans in the surrounding area (watershed and metro-Atlanta counties).

46.1% of Salmonella isolates have PFGE patterns that are indistinguishable from clinical isolates

Several Salmonella strains with the same PFGE patterns were simultaneously recovered from surface water and humans in the surrounding area



On this map we can take a close look at those AR Salmonella as well as one S. Thompson I want to tell you about.

We got all of the MDR Oranienburg from a single sampling event; Shown in red, the majority of those came from sampling a single stream, McNutt's Creek, that is often known to be contaminated with Salmonella and high CFUs for E. coli.

There is a set of chicken houses upstream, but further upstream there are also cattle farms, additionally the same clone is detected at other unconnected streams, so no conclusions can be drawn.

As an example of Salmonella with matching patterns found in both our water samples and in humans at the same time, let's look at Salmonella Thompson, shown in green.

Its PFGE pattern was indistinguishable from one that caused an outbreak in Atlanta at the same time.

However, that outbreak was associated with a Greek restaurant.

It's possible a victim shed that bacterium into the water shed through leaky septic systems, however this is difficult to determine without epidemiological data that was not collected during the outbreak.

PFGE patterns of human isolates are indistinguishable from PFGE patterns of isolates from surface water





We not only isolated ARB from surface water to characterize the bacteria that are culturable and express their ARGs, but also quantified the total ARGs present within the whole bacterial populations in the watershed, including those that have not been expressed.

In addition, antibiotics that are important to human and veterinary medicine were measured to investigate the occurrence and distribution of these antibiotics in aquatic environments.

As WWTPs have been proposed to be the hotspots for the emergence of ARB and from where these bacteria are spread into the natural environment, influents and effluents from three WWTPs located within the watershed and whose effluents flow into the streams within the watershed were included in the analyses.

	No. of water samples positive for AR genes (%)						
AR genes	Fall (n = 38)	Winter $(n = 38)^a$	Spring $(n = 34)^a$	Summer (n = 40			
ermB	23 (60.5)	8 (21.6)	8 (23.5)	11 (27.5)			
tetB	10 (26.3)	2 (5.6)	1 (2.9)	2 (5.0)			
bla _{KPC}	9 (23.7)	5 (13.5)	2 (5.9)	7 (17.5)			
<i>bla</i> _{SHV}	9 (23.7)	2 (5.4)	1 (2.9)	2 (5.0)			
qnrS	8 (21.1)	8 (21.1)	3 (9.4)	8 (20.0)			
bla _{CTX-M}	3 (7.9)	0 (0.0)	0 (0.0)	0 (0.0)			

The most frequently detected resistance gene was *erm*B, which was present in 33.3% (50/150) of the total water samples tested. This was followed by *qnr*S (18.2%; 27/148), *bla*_{KPC} (15.4%; 23/149), *tet*B (10.1%; 15/148), *bla*_{SHV} (9.4%; 14/149) and *bla*_{CTX-M} (2.0%; 3/148).

WWTP samples are not presented because of low no. of samples tested: 9 influent and 6 effluent samples. All influent samples were positive for every AR gene, while a few effluent samples were negative for blaCTX-M and tetB genes.

AR genes	Eall $(n - 38)$ Winter $(n - 38)$ Spring $(n - 34)$ Summer $(n - 34)$						(10)	
	Fall (n	= 38)	Winter	(n = 38)	Spring (n = 34)	Summer	(n = 40)
ermB	1 522 0		255.0	11.2	41.0	average	247.5	11.0
tetB	1,533.8	57.2	355.0	11.5	41.0	1.9	347.5	0.1
blawec	127.0	4.0	40.0	0.0	1.0	0.0	1.0	0.1
blasuv	13.5	1.0	49.9	2.0	70.2	2.4	577.5	9.0
anrS	325.0	10.7	2.0 592.6	0.1 10 <i>5</i>	2.0	0.1	702.0	0.2
bla _{CTX-M}	455.2	12.5	0.0	0.0	0.0	4.0	0.0	0.0

While ARG copy numbers ranged from 10^{0} to 10^{3} copies/mL in surface water, the copy numbers ranged from 10^{3} to 10^{5} copies/mL in influents and 10^{0} to 10^{5} copies/mL in effluents.

qPCR data on ARGs and source tracking genes

- Do ARGs in surface water correlate with source poultry, cattle, or human source tracking genes?
- Do ARGs contaminating surface water correlate with land use?
- Do ARGs and source tracking genes contaminating surface water correlate with the sanitary sewer or septic systems?



This is a comparison of AR genes qPCR with source tracking qPCR for the whole watershed except for the Brooklyn creek samples.

You can see that human fecal genes correlate with AR genes strongly with a rho of .613.

Poultry and cattle fecal genes weakly correlate with AR genes.

Therefore, AR genes are associated with human feces more than poultry or cattle.



We removed Brooklyn Creck from the data on the previous slide, because we were afraid it would skew the results for the whole watershed.

Brooklyn Creek is in a residential area, and you will recognize it and remember it's the same area we isolated that ST131 with the ESBL, because you can seen that Athens Regional is up here, and Saint Mary's hospital is in the middle.

Here the association between AR genes and human fecal markers is very high with a rho of 0.759.

In fact the association is so strong we can look at specific genes, with ermB having a rho of 0.824.

Others include tetB, KPC, SHV qnrS, and CTX-M.

So it appears that human waste is strongly associated with AR genes in the watershed, especially in these residential areas.



We tried to correlate the data with land use, and really got nothing significant. However, if we compared human fecal genes with sewer density we got a strong association with a rho of 0.632.

If we compare AR genes to sewer density we also get an association but it is weaker.

When we compare these markers to average septic age, we get a very weak association.

Therefore, much of the surface water contamination with human markers and AR genes is likely due to our aging and leaking sewer system.

And this is a problem for most cities in America due to the lack of maintenance of our public infrastructure.



Non-point source fecal contamination from aging wastewater infrastructure is a primary driver of antibiotic resistance in surface waters



We not only isolated ARB from surface water to characterize the bacteria that are culturable and express their ARGs, but also quantified the total ARGs present within the whole bacterial populations in the watershed, including those that have not been expressed.

In addition, antibiotics that are important to human and veterinary medicine were measured to investigate the occurrence and distribution of these antibiotics in aquatic environments.

As WWTPs have been proposed to be the hotspots for the emergence of ARB and from where these bacteria are spread into the natural environment, influents and effluents from three WWTPs located within the watershed and whose effluents flow into the streams within the watershed were included in the analyses.



In order to find the source of the AR contaminants in the surface water, we chose WWTPs. We collected influent and effluent samples from 3 WWTPs located within the watershed and whose effluents flow into the streams within the watershed. And we investigated whether WWTPs were effective in reducing AR contaminants and whether WWTPs contributed to the levels of AR contaminants in surface water.







If you look at the upper figures to look at AR gene copy number ratio between wastewater influent and effluent samples, wastewater treatment greatly reduced the absolute copy numbers of ARGs but did not significantly change their relative copy numbers, which was normalized to 16S rRNA gene copy numbers. The absolute abundance of ARGs was potentially decreased due to a reduction in the overall abundance of bacterial populations during the treatment process. However, high density of bacteria and nutrients as well as antibiotics within the treatment system could have led to a favorable environment for horizontal gene transfer of ARGs and therefore a smaller reduction in the relative abundance of ARGs.

But when the antibiotic concentrations in the influent and effluent samples were compared as shown in the bottom figure, WWTPs did not remove antibiotics as efficiently as ARGs. The average antibiotic concentration decreased by half, but this difference was driven by small decreases in most of the antibiotics, although most of these decreases were not statistically significant on an individual basis except for lincomycin.

WWTPs as a source of AR contaminants? When WWTP influent samples and effluent samples were compared: WWTPs only partially removed AR bacteria, AR genes, and antibiotics, with the effluents containing high levels of AR contaminants WWTPs were not very effective in removing AR contaminants, releasing the contaminants into receiving water When water samples collected upstream and downstream of WWTP were compared: No significant differences in AD explored to be two on the second to the second to

• No significant differences in AR contaminants between the upstream samples and downstream samples

But when we compared the water samples collected upstream and downstream of a WWTP, there was no significant difference in AR contaminants between the upstream samples and downstream samples.

Also, only a small number of the water sampling sites received wastewater effluents, but most of the sites contained high levels of AR contaminants throughout the year, indicating that there are other sources of AR pollution apart from WWTPs.

So our conclusion was that although WWTPs contribute to the AR contamination in surface water, they are not the main source of AR in surface water of the Upper Oconee Watershed.



This paper on the presence and distribution of AR bacteria, AR genes, and antibiotics in surface water and the impact of WWTPs on AR in surface water has been just accepted for publication. So, if you are interested, you could learn more about this in this paper: Distribution of Antibiotic Resistance in a Mixed-Use Watershed and the Impact of Wastewater Treatment Plants on Antibiotic Resistance in Surface Water.

How is this surveillance data used in One Health approach?

- The National Antimicrobial Resistance Monitoring System has no means to integrate this data
 - Where did the Salmonella in the water come from?
 - Agriculture is the assumed source, but sewers leak and wild animals poop, both can be a source of *Salmonella*
 - What is the risk that *Salmonella* found in surface water will contaminate food and infect a human?
- Current efforts to reduce Salmonella infections
 - USDA Food Safety Inspection Service: eliminate Salmonella from U.S. poultry trough regulations



Salmonella causes about 1.3 Million infections each year, most of which are caused by specific serotypes.

The Healthy People 2030 goal is to reduce human infections by 25%, while poultry only causes \sim 21% of cases as seen on the previous slides.

We have been trying to reduce Salmonella infections for the past three decades, and the goals of reduction of Salmonella in retail poultry products have been successful, but the levels of Salmonella infections in humans have not decreased.



Salmonella is a gram-negative, rod-shaped bacteria, that is highly diverse and includes over 2,600 serotypes.

According to FDA, Salmonella infections that could be associated with food fell into these percentages.

Meat, fish, seafood, and other meat accounted for roughly 54% of attributed human infections in the U.S.

About 21% of human infections were associated with poultry.

Turkeys, shown in yellow, account for 7% and chickens, also shown in yellow accounting for 14% of Salmonella infections.

It is interesting to note that almost half of Salmonella infections are associated with products other than meat.

And we need to keep in mind that attribution is difficult and usually relies on a large enough outbreak to trigger an investigation, while most Salmonella infections are sporadic and never linked to a source.

Therefore, these numbers may not represent the true effect of the different commodities on human Salmonella infections.



- The persistence of salmonella illness is occurring despite less contamination
 of meat
- The posted framework under consideration, uses three strategies to target *Salmonella* at different points in the slaughter and processing operation.
 - Testing for Salmonella before entering an establishment.
 - Enhancing establishment process control monitoring and FSIS verification.
 - Implementing an enforceable final product standard that includes serotype and quantification.
 - So our conversation today about what interventions work best and what indictors are most useful is very timely.



This graph shows a seasonal trend in Salmonella infections.

Y-axis represents the number of Salmonella infections reported to CDC and X-axis represents the month the specimens were collected.

There is a clear pattern repeating every year with more infections during summer months.

Why would a food borne disease be seasonal?

There simply can't be that much potato salad, plus, most cases reported are not associated with an outbreak and are what we call sporadic cases with no known source.

So what are some other reasons summer months it goes from less than 400 to nearly three times that?

Other things people do in the warm months is outdoor recreation, and a lot of that has contact with surface water, including swimming, fishing, boating, etc.

So, why not look for the cause of this increase in the environment?

46.1% of Salmonella isolates have PFGE patterns that are indistinguishable from clinical isolates

Several Salmonella strains with the same PFGE patterns were simultaneously recovered from surface water and humans in the surrounding area



Gabi and lab did it all with lots of help from other BEAR members!

