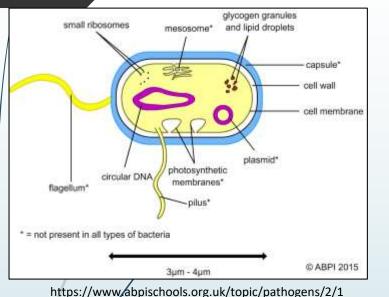
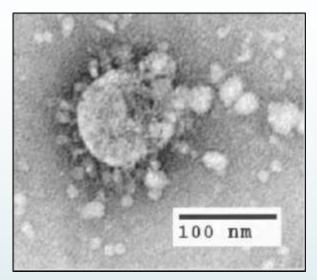
Signal Optimization and Chemometric Analysis of Laser-Induced Breakdown Spectroscopy Bacterial Spectra to Quantify Detection Limits and Improve Classification Accuracy

> Jeremy Marvin Department of Physics University of Windsor

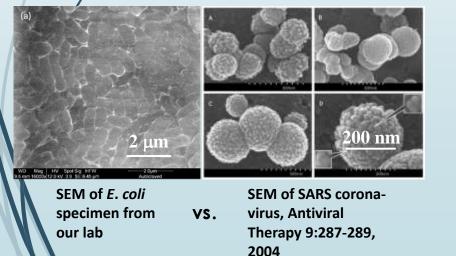
LIBS on Viruses? Size matters!



- Bacteria are ~1-3 µm
- Corona viruses are ~100-300 nm
- Volume is roughly 1,000 10,000 lower!
- Also, viruses are not rich / don't contain trace metals, as bacteria do.



C.S. Goldsmith, CDC, https://www.cdc.gov/sars/lab/images.html



Two known papers on the use of LIBS to identify viruses: (full details in S.J. Rehse, Spectrochimica Acta Part B 154 (2019) 50–69)

detect the presences of an MS-2 bacteriophage (smallpox surrogate) J.L. Gottfried, Anal. Bioanal. Chem. 400 (2011) 3289–3301,

differentiation with LIBS of four strains of live *hantavirus* R.A. Multari et al., Appl. Opt. 51 (2012) B57–B64,

Motivation

- Current methods of bacterial identification in a clinical setting
 - require transferring the sample to a lab
 - require expertise in microbiology
 - expensive/labor-intensive
 - may only be useful for certain types of bacteria
 - slow

For example: standard culturing techniques for bacterial identification take 1-3 days

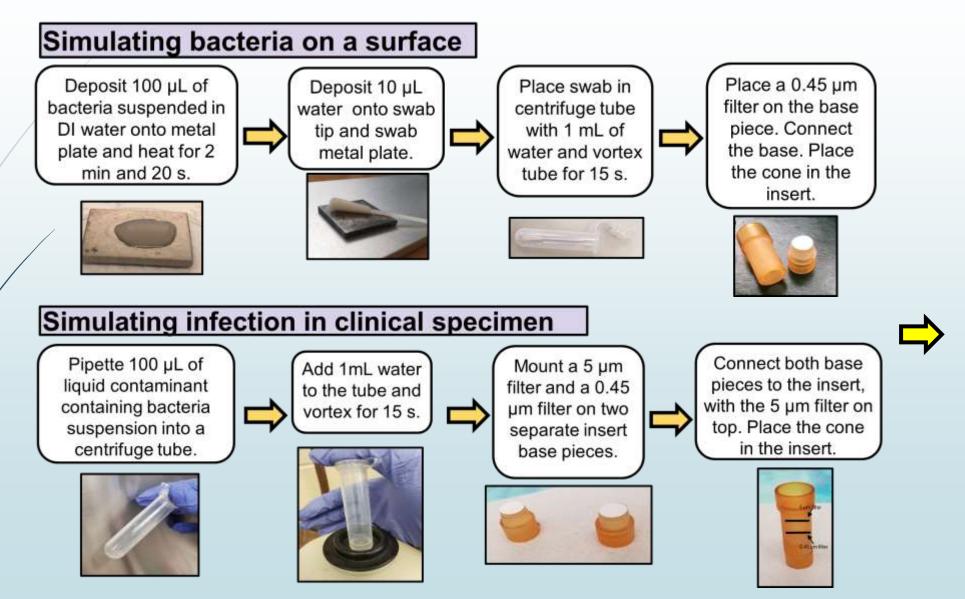
- Patients are treated with broad-spectrum drugs that have given rise to the crisis of antibiotic resistant bacteria
- Rapid and accurate diagnosis of bacterial infection are required so that more targeted treatment can begin as soon as possible

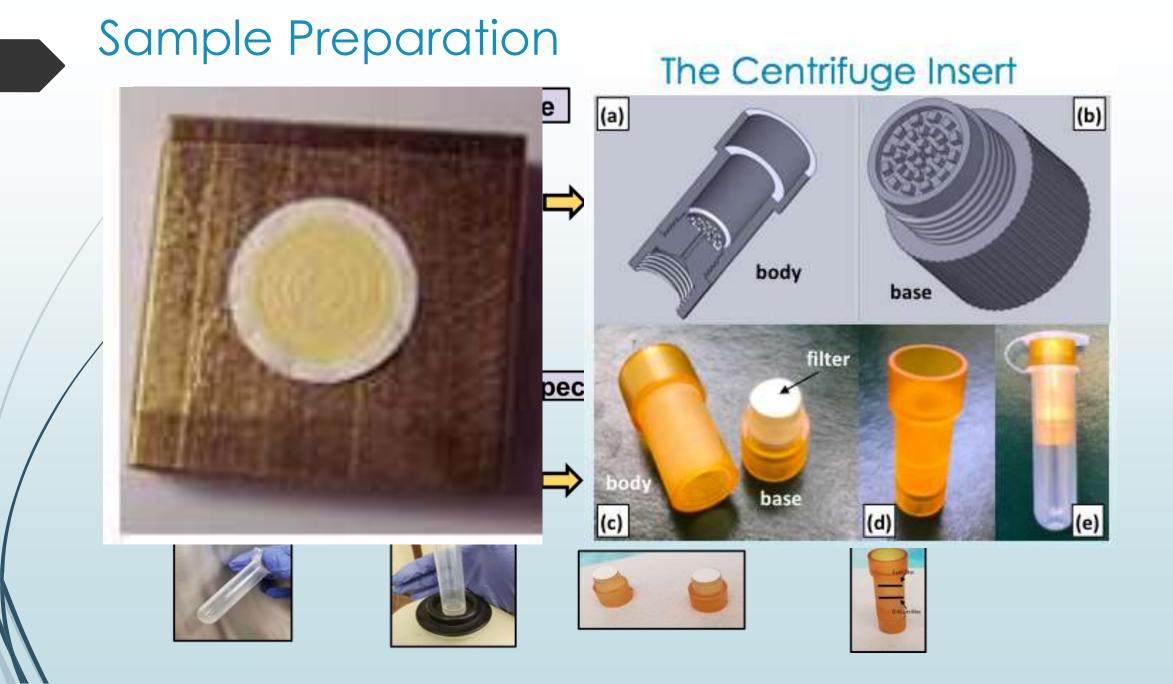
Goal

Develop LIBS as a rapid point-of-care diagnostic tool in a clinical setting

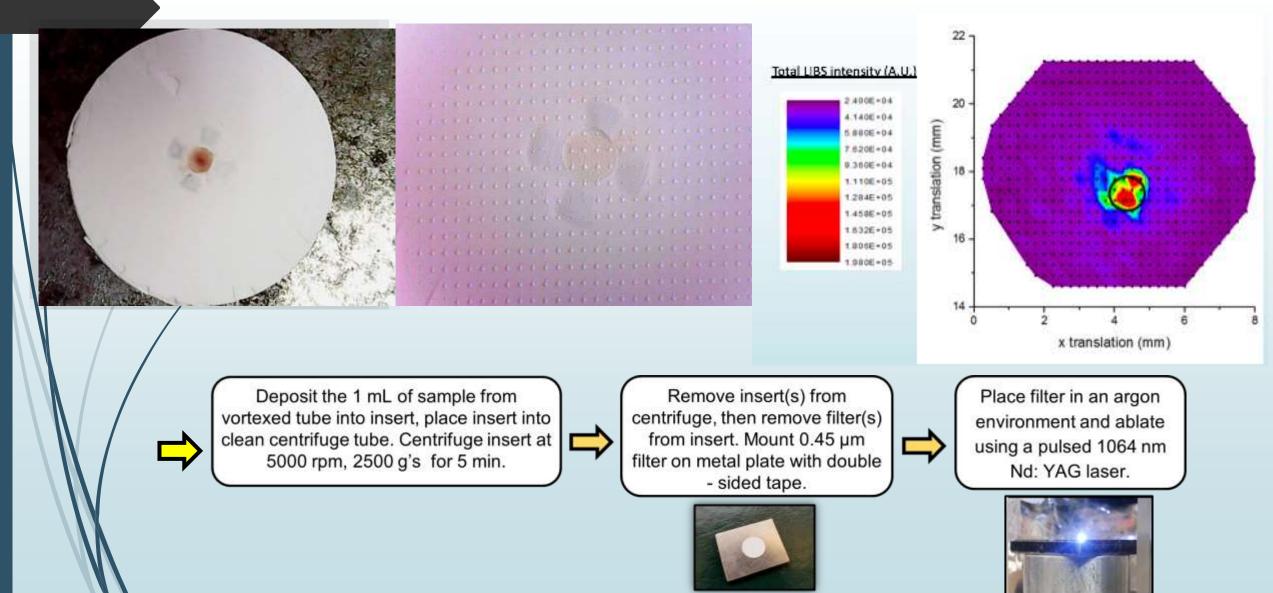
This includes developing quick bacterial preparation methods prior to testing that utilizes equipment and methods that are common or easy to implement in a clinical setting

Sample Preparation

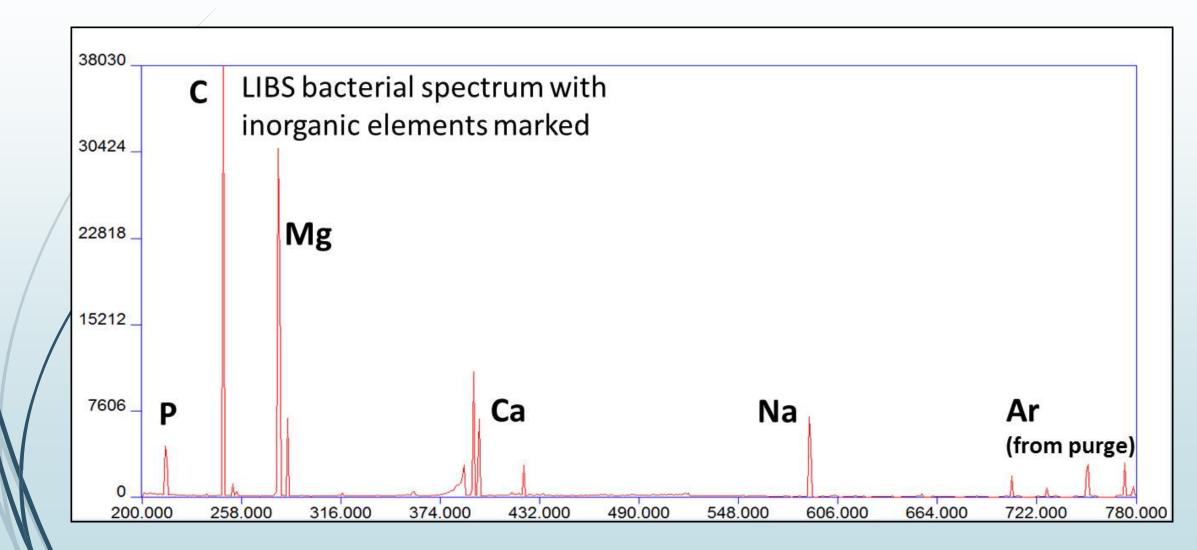




Concentrating and Shooting Bacteria on Filters



Spectra

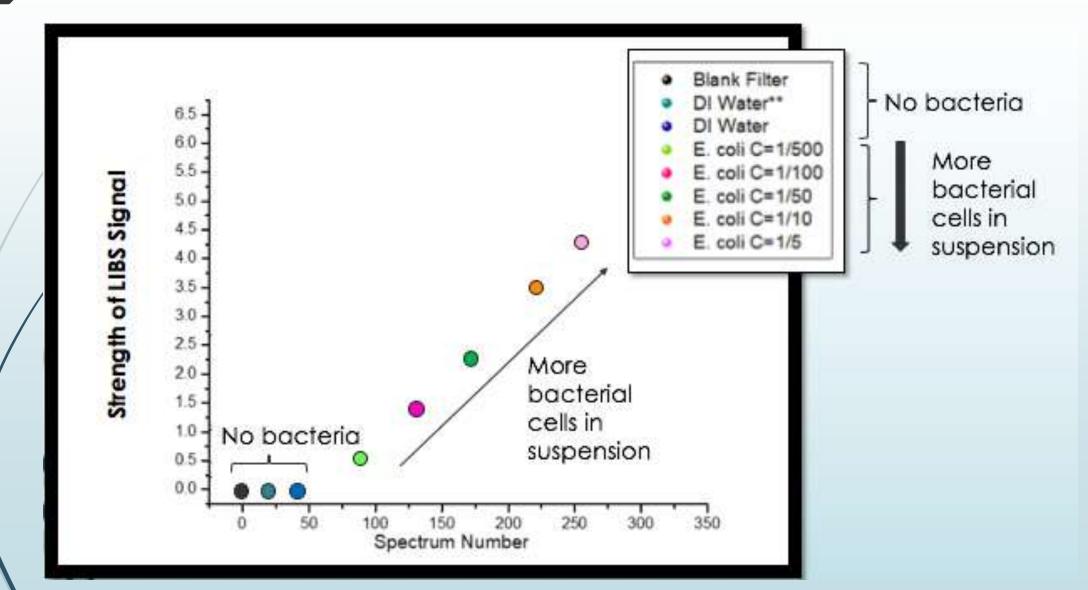


Focus of our Work

- The intensities of emission lines in the LIBS spectrum provide a unique elemental spectral fingerprint for each type of bacteria which can be classified using chemometric algorithms.
- Our goals:
 - accurately identify/classify as small a number of bacterial cells as possible (lower LOD)
 - maximize the rates of true positives while minimizing the rates of false positives.
- How can we improve classification accuracy?
 - reducing background signal in the spectrum
 - investigating data pre-processing, and the
 - use of silver microparticles to produce a more repeatable and robust spectrum.

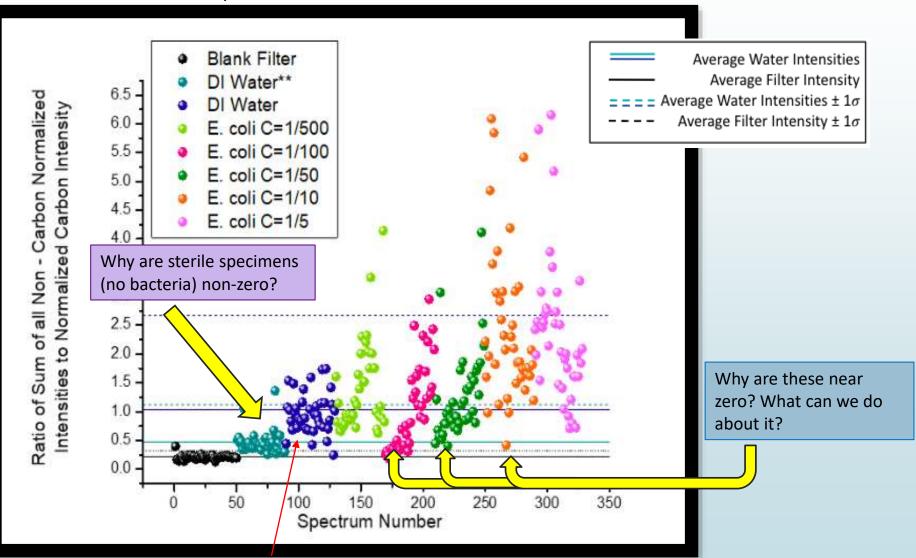
Our Expected Curve-of-Growth

A LIBS bacterial curve of growth **<u>SHOULD</u>** look something like this



Our "Confusing" Curve-of-Growth

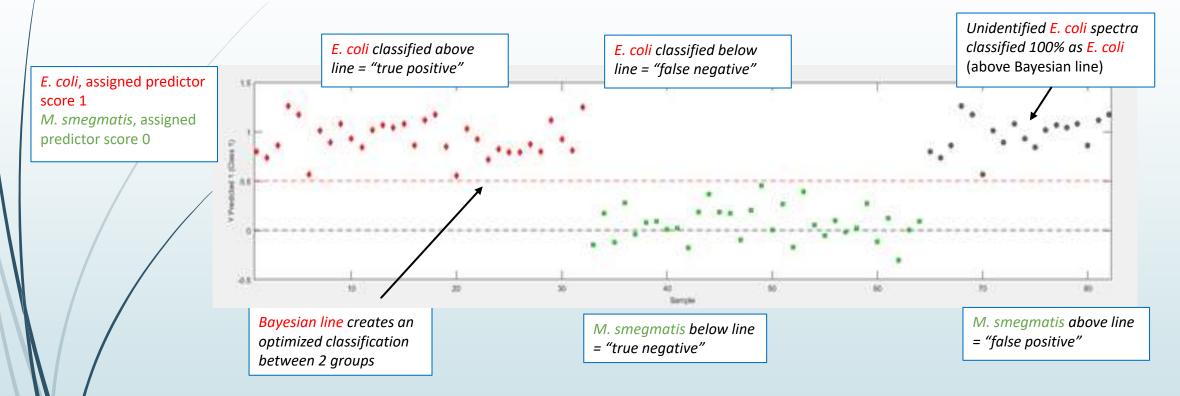
Our LIBS bacterial curve of growth (constructed with serial dilutions of *E. coli* and sterile water) looks like this...



"Rigorous cleaning" has eliminated these particular strong blue water data.

Chemometric Analysis

An example of a PLSDA discrimination between E.coli and M. smegmatis

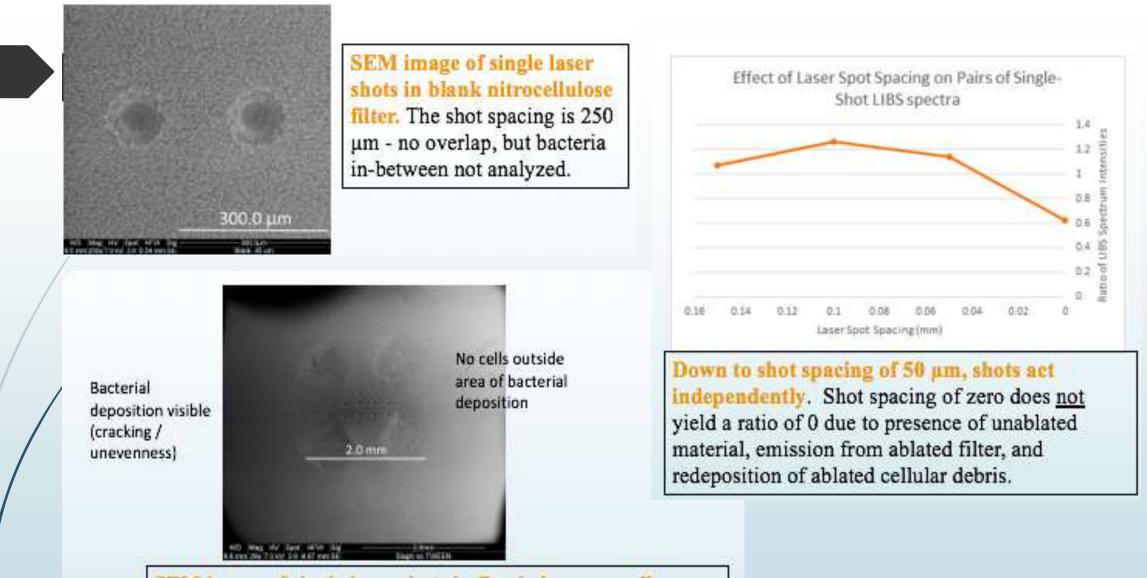


- 164 unique and independent variables are created from 18 emission lines.
- Classified by DFA or by PLSDA.
- Both techniques result in a sensitivity and a specificity
- Improve accuracy by training a library through objective data rejection techniques

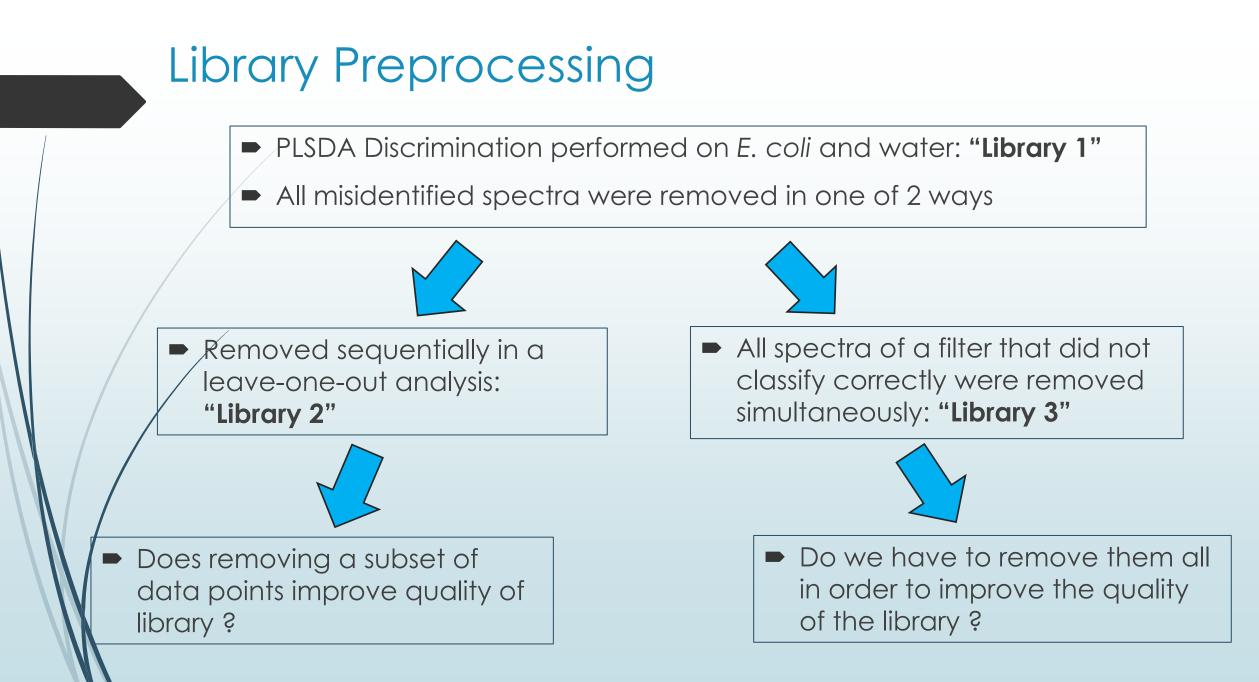
Current Total Library

To classify bacteria using chemometrics, an extensive collection of "known" spectra with hundreds of data points is required. This is called a "library."

	Current Total Library	Total	
	Escherichia coli (180)		
eria	Mycobacterium smegmatis (150)	653	
Bacteria	Enterobacter cloacae (113)		
ñ	Pseudomonas aeruginosa (100)		
	Staphylococcus aureus (110)		
ank	DI Water (260)	510	
B	Nitrocellulose Filter (250)	510	



SEM image of single laser shots in *Staphylococcus* cells deposited in circular area 900 µm in diameter by centrifuge cone. One shot of "blank filter" also obtained.



	External Validation of DI Water	Average Sensitivity
	Library 1	78.40%
	Library 2	78.40%
/	Library 3	75.90%
	External Validation of E. coli	Average Sensitivity
	Library 1	72.50%
	Library 2	88.80%
	Library 3	88.80%

- There was no significant improvement upon removing spectra of DI water that classified incorrectly.
- There was some improvement for individual data sets of E. coli, however those data sets that improved markedly had the majority of spectra removed.

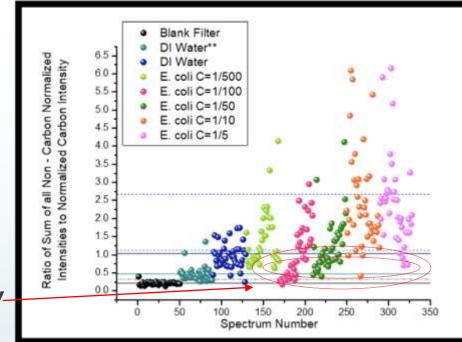
Outlier Rejection

Two tests were investigated to identify outliers:

- Method 1: Water Threshold Analysis
 - Spectra were excluded if their intensity was consistent with the average water signal intensity +/- 1σ displayed below in red of tables.

Outlier Rejection Method	Sensitivity (CV)	Specificity (CV)
Unprocessed	97.5%	100.0%
Water ± 1σ	94.4%	100.0%

C = 1/5 dilutions only



Outlier Rejection Method	Sensitivity (CV)	Specificity (CV)
Unprocessed	85.5%	87.2%
Water ± 1σ	67.8%	79.0%

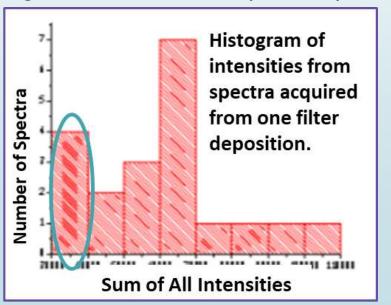
All bacterial concentrations

Outlier Rejection

- Method 2: Histogram Analysis
 - All the spectra in the bin containing the weakest intensities were taken to represent 'empty shots' and were removed from the library.
 - The binning was chosen automatically.
 - Results shown in gold of the table

Outlier Rejection Method	Sensitivity (CV)	Specificity (CV)	
Unprocessed	97.5%	100.0%	
Water ± 1σ	94.4%	100.0%	
Histogram	100.0%	96.9%	

*Still analyzing this method when applied to all concentrations of data

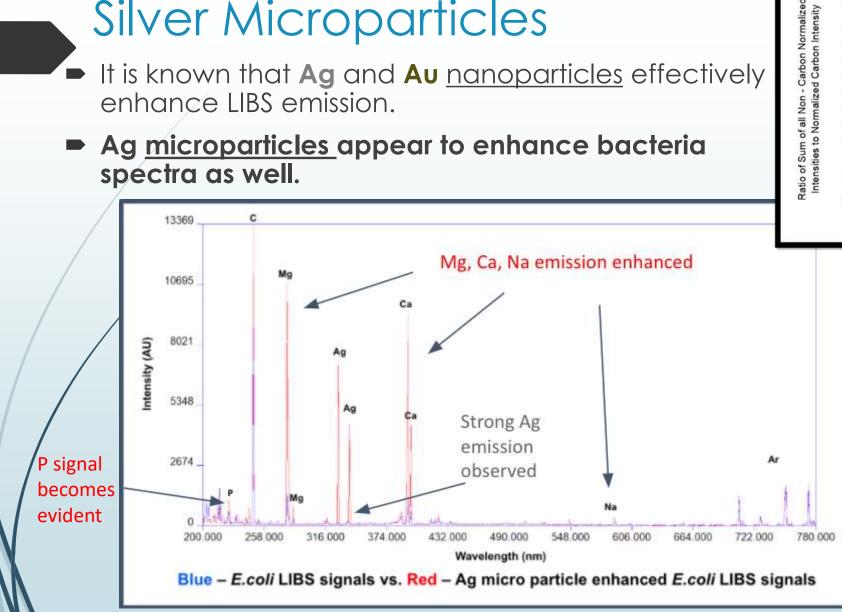


Histogram of intensities from spectra acquired from one filter deposition.

The column circled represents the 'empty shots' which clearly do not follow a normal distribution for bacterial spectra. In this case, 4 of 23 spectra were rejected.

Silver Microparticles

- It is known that Ag and Au <u>nanoparticles</u> effectively enhance LIBS emission.
- Ag microparticles appear to enhance bacteria spectra as well.



If we can't eliminate them with outlier rejection, maybe we can stop these seemingly blank bacteria data from ever occurring using enhancement.

250

Blank Filter

E. coli C=1/5

3.5

3.0

20.

coli C=1/500 C = 1/100

100

150

Spectrum

200

*Note: this is not the same enhancement that was mentioned in the nanoparticle talk yesterday

Preliminary Data – Currently Under Investigation

Average Elemental Enhancement of 3 Bacteria Species with the Addition of Silver Microparticles

	с	Р	Mg	Ca	Na
Enhancement of <i>E. coli</i>	1.3	4.6	3.9	5.3	3.9
Enhancement of <i>M. smegmatis</i>	1.1	1.1	0.8	1.9	2.1
Enhancement of <i>P. aeruginosa</i>	1.3	1.1	6.9	27.3	1.0



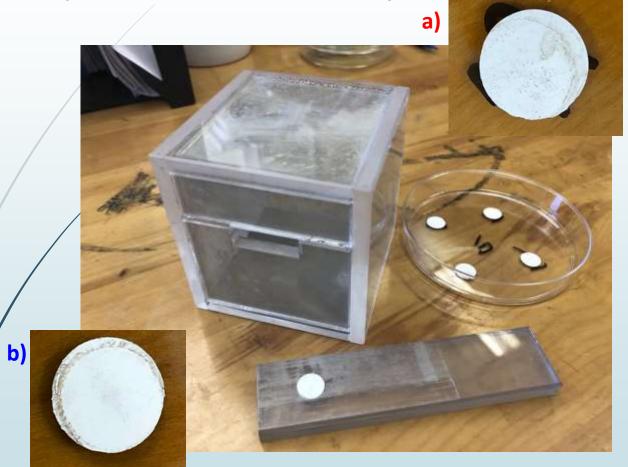
https://www.reade.com/products/silver-powder-silver-metal

Not all elements are enhanced in the same way

Spectra appear to be stronger now

Question: Could this eliminate empty spectra from occurring and improve our overall limit of detection?

A **custom sealed chamber** was built to agitate the silver micro-powder. Filters inserted into the chamber collect trace powder as it settles. The amount of silver, shaking, and settling time were adjusted to obtain a uniform coverage.



Two methods for filter preparation.

0.5 - 1 micron spherical silver
(99.9%) powder:
a) Spread on filter (without chamber) vs
b) Trace uniform spread (with chamber)

Next Steps: Quantify enhancement in terms of surface coverage of silver microparticles using mass, density, and diameter of filter media. SEM images to confirm coverage. Test gold microparticles and nanoparticles.

The people who did the work...



NSERC Discovery Grant

*

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100 100 100 200 200 20

Natural Sciences and Engineering Research Council of Canada

Conseil de recherches en sciences naturelles et en génie du Canada

CFI-LOF grant



University of Windsor

