







Homeland Defense & Security Information Analysis Center

IARPA's SILMARILS: Active IR Spectroscopy for Chemical Detection

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Intelligence Advanced Research Projects Activity (IARPA)



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The views presented are those of the speaker and do not necessarily represent the views of DoD or its components.

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Introduction

HDIAC & Today's Topic



HDIAC Overview

What is the Homeland Defense & Security Information Analysis Center (HDIAC)?

One of three Department of Defense Information Analysis Centers

Responsible for acquiring, analyzing, and disseminating relevant scientific and technical information, in each of its eight focus areas, in support of the DoD and U.S. government R&D activities

HDIAC's Mission

Our mission is to be the go-to R&D/S&T and RDT&E leader within the homeland defense and security (HDS) community, by providing timely and relevant information, superior technical solutions, and quality products to the DoD and HDS Communities of Interest/Communities of Practice.



HDIAC Overview

HDIAC Subject Matter Expert (SME) Network

HDIAC SMEs are experts in their field(s), and, typically, have been published in technical journals and publications.

SMEs are involved in a variety of HDIAC activities

- Authoring HDIAC Journal articles
- Answering HDIAC Technical Inquiries
- Engaging in active discussions in the HDIAC community
- Assisting with HDIAC Core Analysis Tasks
- Presenting webinars

If you are interested in applying to become a SME, please visit HDIAC.org or email info@hdiac.org.





Dr. Kristy DeWitt

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Academic

- Virginia General Equivalency Diploma, 1996
- B.S. in Chemistry & Physics, minor in Mathematics, Mary Washington College, 2000
- Ph.D. Chemistry, University of Virginia, 2005
- National Research Council (NRC) Post-Doc, National Institute of Standards and Technology (NIST), 2005-2007

Professional

- IARPA Program Manager focusing in chemical detection (SILMARILS, MAEGLIN, and Ithildin programs, various seedlings)
- Previous work as a technical SETA contractor at IARPA and DARPA
- Also worked on the performer work at several companies developing fiber lasers, wind sensing systems, and skin detection sensors



Overview



Office of the Director of National Intelligence

Central Intelligence Agency

Defense Intelligence Agency

Department of State







National Security Agency

Department of Energy





National Geospatial-Intelligence Agency

Department of the Treasury







National Reconnaissance Office







Department of the Army

Federal Bureau of Investigation





Department of the Navy

Department of **Homeland Security**







Air Force

Coast Guard

Marine Corps



IARPA Mission & Method

IARPA envisions and leads high-risk, high-payoff research that delivers innovative technology for future overwhelming intelligence advantage

- Our problems are complex and multidisciplinary
- We emphasize technical excellence & technical truth

Bring the best minds to bear on our problems

- Full and open competition to the greatest possible extent
- World-class, rotational Program Managers

Define and execute research programs that:

- Have goals that are clear, measureable, ambitious and credible
- Employ independent and rigorous Test & Evaluation
- Involve IC partners from start to finish
- Run from three to five years
- Publish peer-reviewed results and data, to the greatest possible extent
- Transition new capabilities to intelligence community partners



IARPA Chemical Detection Efforts: A Technology Toolkit Tailored to the IC's Needs

Mission Space

- Denied Territory
- Homeland Security
- Forensic Analysis

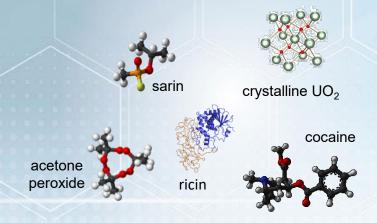
Compounds of Interest

- Chemical weapons/poisons
- Explosives (primary, secondary, oxidizer/fuel, HMEs)
- Narcotics (illegal, prescription, designer)
- Nuclear fuel cycle material
- Bio-weapon signatures

Collection/Analysis modalities

- Trace surface residue
- In situ gas phase collection/analysis
- Field collection for lab analysis

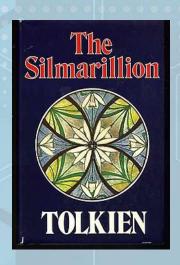






What's in a Name?







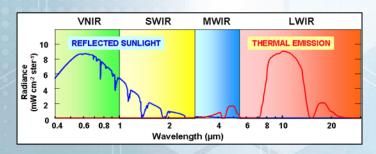
In J.R.R. Tolkien's mythology, the Silmarils are the three jewels crafted by Feanor which hold the unmarred light of the two trees. It is light from a Silmaril that Galadriel gives Frodo in the first Lord of the Rings movie in a glass vial to have as "a light in dark places where all other lights have gone out". Morgoth was the evil master-Lord who stole the Silmarils from the elves, and wore them mounted in an iron crown. Many wars were fought, and lives lost over possession of the Silmarils.



SILMARILS Motivation

Current state-of-the-art in surface chemical detection:

- Passive hyperspectral
 - Ambient light/thermal emission as source
 - At least 3-10% of surface area must be covered by target chemical (mg/cm² to g/cm²)
- Handheld IR/Raman sampler
 - Physically collect sample from surface & insert in sampler
 - Limited throughput, overt, must physically access sample, limited library
- Mobile / National lab
 - Physically collect sample, bring to lab
 - Limited throughput and timeliness



USGS mineral mapping of Afghanistan with HyMap SWIR HSI sensor to support crop options other than poppy





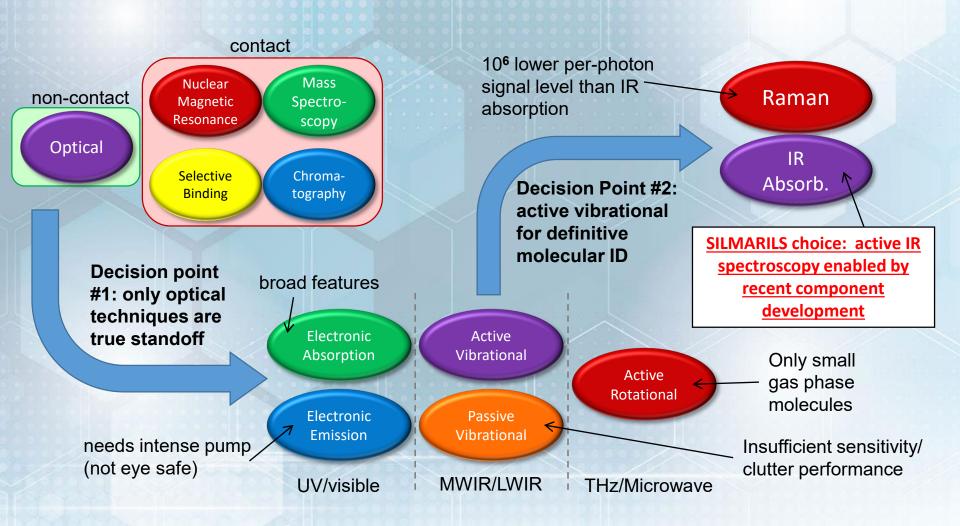
Handheld IR/Raman sensors



Capability Gap: High throughput ID of trace surface chemicals in clutter



Developing an Approach: Start at the Beginning





Program Overview

- Standoff ILuminator for Measuring
 Absorbance and Reflectance Infrared
 Light Signatures
- Prototype portable scanner for field forensics with 30+ m standoff range
- High sensitivity & specificity, eye safe, works with complex mixtures
- Sensitive enough to detect 10th fingerprint after handling explosives / narcotics



Gas	Solution	Solid/Liquid
0.1 ppm	10 μg/mL	1 μg, 0.1 μ/cm ²



Scan people/vehicles/cargo moving through portal or past a monitor station at normal speed



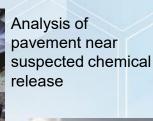
Scan room or vehicle for target chemicals



Explosive residue in fingerprints on car door: parking garage or approaching checkpoint



Monitor factory emissions for transients



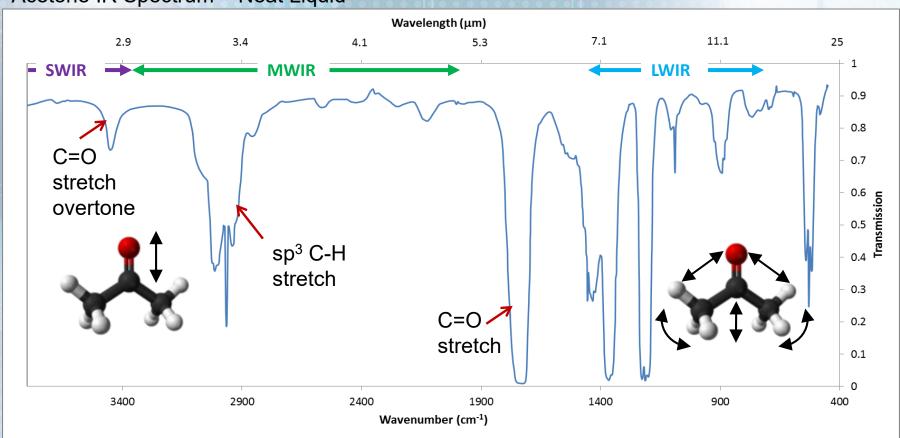


Goal: Laboratory performance in a field instrument



IR Spectrum Provides Molecular Barcode

Acetone IR Spectrum - Neat Liquid



SWIR – combination and overtone bands

<u>MWIR</u> – "functional group" region: Bending and stretching motions of a few directly connected atoms

LWIR – "fingerprint" region: Twisting, rocking, breathing modes of whole molecule



Challenges for Standoff IR Spectroscopy

Part 1: Spectrometry

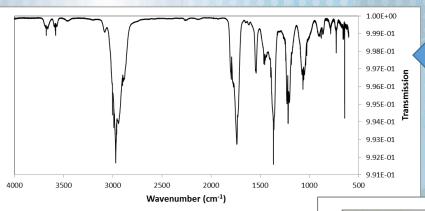
- Current active IR techniques limited by sources & detectors
 - Broadband incoherent light sources require large optics, have limited range
 - Narrow-band coherent sources limit spectral coverage & therefore specificity
 - Spectrometer-based detection trades SNR, resolution, scan speed

Part 2: Spectroscopy

- Target signatures must be distinguishable from spectral features belonging to other compounds in same sample area
 - No atomically clean surfaces or evacuated gas cells
- Signal-to-clutter ratio often more important to detection threshold than signal-to-noise
- "Trace" features do not look like "bulk" features, either have to measure every possible combination, predictively model, or focus on conserved features

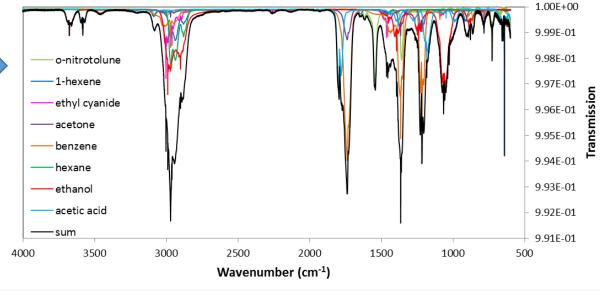


Complex Spectrum Example



Raw transmission spectrum of nine gas phase chemical species at ~ equal concentration

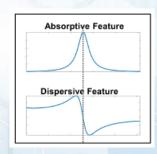
Principal components analysis showing contribution of each species to total spectrum



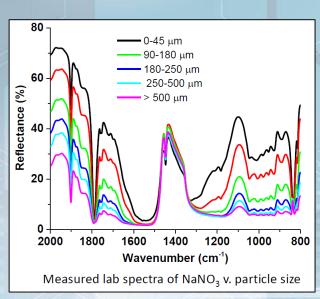


Reflectance Spectra of Solids

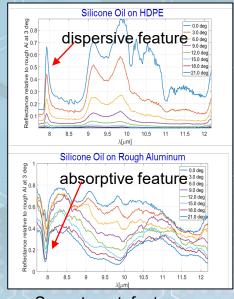
 Feature intensity and even absorbtive/dispersive quality changes based on target size, crystallinity, and substrate interactions, but feature location does not – feature based, vs. traditional "spectral shape" based ID eliminated need for large library of signature permutations for a given target



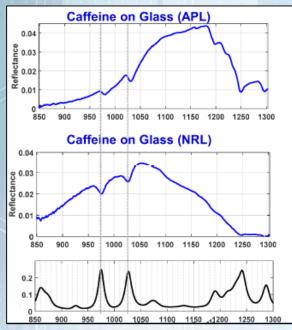
INTRINSIC and **EXTRINSIC** factors



Particle size changes some (but not all) peak intensities



Same target, features absorptive vs. dispersive depending on substrate



Same target and substrate, absorptive features for thin film, dispersive for particles



Program Structure Overview

- original
- added Fall 2017
- added spring 2018

Item	Phase 1	Phase 2	Phase 3
Duration	18 months	18 months	18 months
Scope	Algorithms: Test against Government provided data set. Spectrometer: Laboratory breadboard traceable to Phase 3 design. Performance demonstration against standard samples representative of real-world background/ clutter.	Spectrometer components near final SWaP, but system still laboratory breadboard. Integration of spectrometer and algorithms. Incorporate new spectroscopy/sample methods to improve sensitivity/specificity. Add metrics specific to each performer's key advantages to enable early transition path(s). Participate in outdoor field tests organized by mission partners.	Prototype field demonstration with real-time processing (both spectral reconstruction and chemical detection/ identification algorithms). Require participation in field test exercises held by mission partners.



Performers and Approaches



19700	Performer	System Approach	System Key Features	Algorithm Approach	Algorithm Strengths/ Weaknesses
7 7	Block MEMS	Adaptive step scan	Very long range w/ increased power	Combination of ACE, Sparse ACE, and Bayes	Complimentary approaches, Bayes scales to large libraries; computational complexity
\ \ \ \ \	→ Leidos	Scanned FTIR	Rapid forensic imaging	Combination of Adaboost with array of classifiers for solids and liquids and ACE for gases	Training can find classification "clues not seen by humans"; challenged by large libraries and clutter
	LGS	Dual comb spectroscopy	High resolution gas spectroscopy for CBRN	Decision tree-based machine learning	Can account for a wide variety of spectral variation; computationally intensive for very large libraries
00000	Physical Sciences	Spatial heterodyne	Moving targets	ACE	Mature algorithm; does not scale with library size, requires separate "target" and "background" (non-target) pixels to be fed to it, little innovation over current SOA
\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\	Spectrum Photonics	Thin disk interferometer	Very low cost, low power for shorter range	Single-pixel approach similar to ACE, eigenvalues and matched filters	Uses spatial information, uses filters to minimize influence of noise; cannot handle large libraries with limited spectral bands

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3 Performers in Phase 2

Performer	Approach	Key Feature	
Block MEMS	Rapidly tunable quantum cascade laser source, high resolution hyperspectral imaging	Adaptive laser step-scan focuses on feature-rich areas, high resolution imagery mitigates need for substrate signatures in library, feature-based detection algorithms	Wavelength Tuning Laser Camera Signature Signature Wavelength (µm) Hyperspectral Image Cube Detection Map 10 0 8 8 6 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
Leidos	SWIR/MWIR/LWIR Supercontinuum laser source, rotating disk FTIR	Broad-band source minimizes laser speckle effects, FTIR provides throughput & sensitivity advantages	FIIR Spectrometer Beam Sampler Supercontinuum Source Detector
Spectrum Photonics	Thin disk interferometer directly mounted to SLS focal plane array	Very low cost, low power for shorter range – potential for "cell phone chemical imaging"	"FPA sized" LWIR Interferometer



The Whole Phase 2 Team



Block MEMS

- Systems & Technology Research
- Alpes Lasers



IARPA

Program Manager: Kristy DeWitt



Leidos

- Omni Sciences
- · University of Michigan
- IR Flex Corporation
- Arete Associates



Test and Evaluation (T&E) Team

- Johns Hopkins University Applied Physics Laboratory
- Naval Research Laboratory
- Pacific Northwest National Laboratory
- Sandia National Laboratory



Spectrum Photonics

- Arete Associates
- Zeteo Tech
- University of Hawaii, Manoa
- QmagiQ
- OptX



Contracting

• AFRL



Test & Evaluation: Measuring Progress

T&E team Tasks

- Reflectance (solids) and transmission (liquids and gases) measurements for chemical library, complex index of refraction measurements
- Benchmark Tests at month 6 (performer site) and 12 (Government site) for Leidos and Block MEMS, single Benchmark test (performer site) for Spectrum Photonics at month 9
- Final Exam test at Government site for all performers in month 17
- Monthly deliveries of physical sample coupons (from month 5) and synthetic sample coupons (from month 7), critical evaluation of performer reporting
- SME support of Chemical List and Test Plan development, critical evaluation of performers during monthly telecons, periodic review meetings

Organization	SME Support	Meeting Support	Physical Sample Creation	Synthetic Sample Creation	Synthetic Data Model Development	Compound and Coupon Measurements	System Testing	Final Overall Evaluation
JHU/APL	X	X	X	X			Χ	X
NRL	X		X		X			X
PNNL	X					X		X
SNL	X				X			X



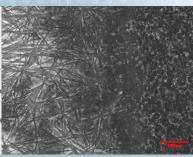
Test and Evaluation Snapshots



Shaped synthetic finger



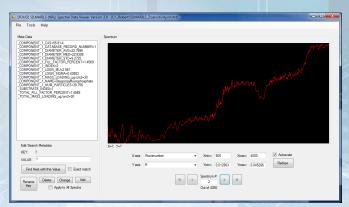
Caffeine on glass, sieved



Caffeine on glass, sprayed

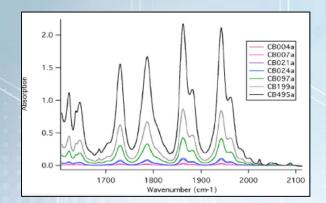


Airbrushed samples at month 6 test site



Synthetic data viewer and spectra

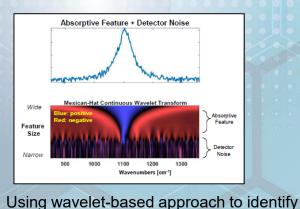




Very short path-length liquid cells to enable full dynamic range of liquid measurement made by squeezing commercial liquid cells in a press

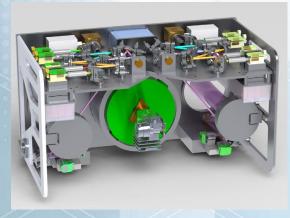


Block MEMS Snapshots



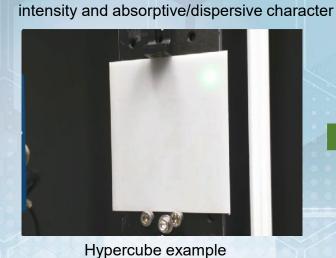
Algorithm team at benchmark test

100



Phase 2 compact breadboard design

Wavelength [µm]



chemicals by feature positions, independent of



0.8 80 P.0 P.0 Reflectance Reflectance 0.0 7.0 y [mm] 60 0.2 0.2 0 1100 1200 20 40 60 80 100 x [mm] Wavenumbers [cm⁻¹]

0.8

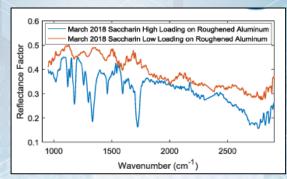
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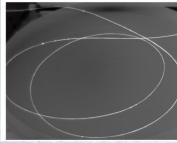
Leidos Snapshots



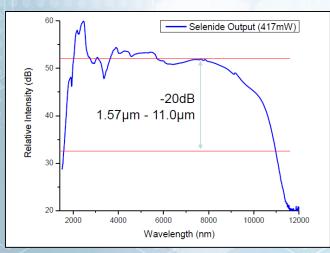
Breadboard testing in April 2018



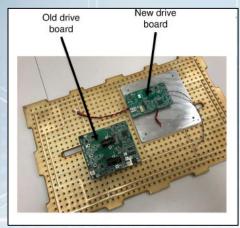
Testing monthly sample coupons



Bright spots in chalcogenide fiber show inclusions/failure points



LWIR source output

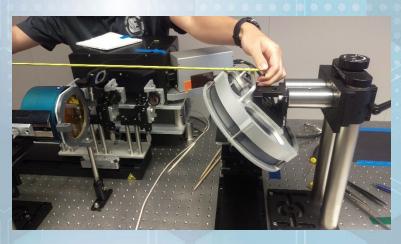


New electronics save SWAP

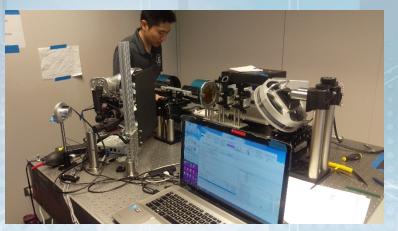
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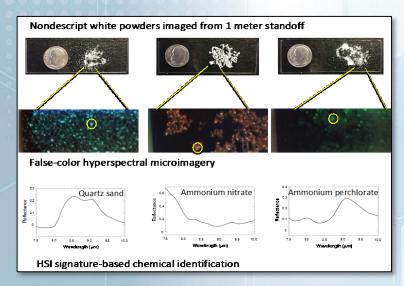


Spectrum Photonics Snapshots

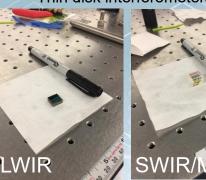


Breadboard testing in July 2018





Thin disk interferometers





Single frame interferogram of ~4 micron single wavelength source

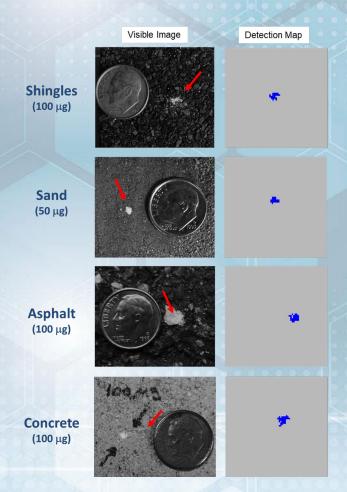


Example Results

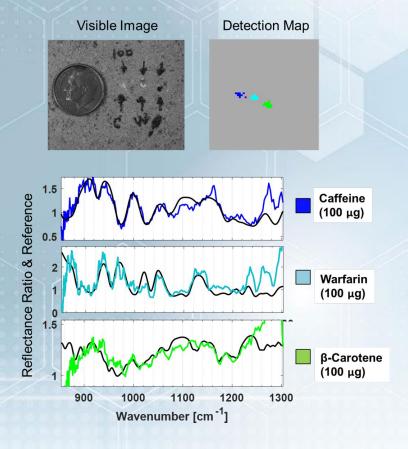


5 Meter Standoff

Caffeine on Outdoor Surfaces



Multiple Powders on Concrete



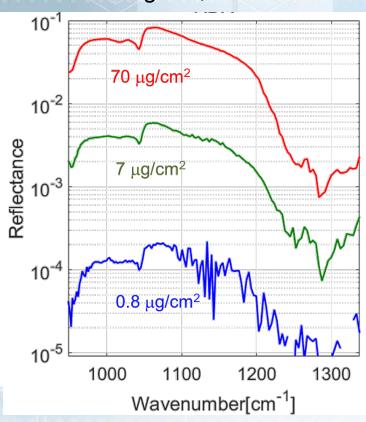


1 Meter Standoff

Caffeine on stone shingles, ~1m standoff

Visible Image **Detection Score Detection Map** Caffeine reflectance, R_{caffeine} Substrate reflectance, R_{substrate} 0 └─ 850 1100 1150 1200 1250 Reflectance Ratio R_{caffeine}/R_{substrate} Library spectrum for caffeine 850 1200 1250 Wavenumber [cm⁻¹]

RDX on glass, 1m standoff



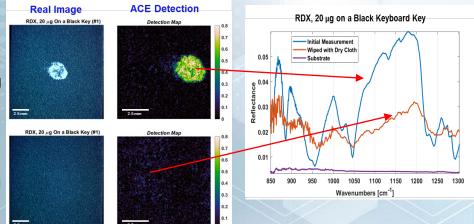


Explosives on Portable Electronics

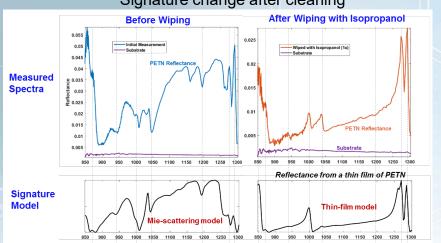
- Determine ability of SILMARILS to detect trace explosives residue on materials associated with portable electronic devices
 - Tested: RDX & PETN
 - 8 different material samples from used portable electronics
 - RDX limit of detection 3 ng, PETN 6 ng
 - Compatible with ETD sensitivity levels
 - Contamination still detectable after cleaning with dry cloth, water, isopropyl alcohol
 - IR signature changes after cleaning indicates that remaining contaminant is very small particles or thin film

Before cleaning

After cleaning



Signature change after cleaning





Packaging, Screening, Forensics

- Wide variety of material texture, porosity, and reflectance
- "Wild" real-world substrates, no cleaning prior to deposition (tires from scrap lot, fabrics washed & worn, cardboard box delivered by Amazon, Coke can from recycle bin, etc.
- Deposition of 50μg, 10μg, and 1μg quantities by dry transfer

Polyurethane vinyl	Duct tape	Slate	Polished Al
Packing tape	Glass tile	ABS plastic	2x4 (pine)
Stainless steel	Steel belted tire	Hardwood (cherry)	Drywall (painted)
Cardboard	Marble	Canvas	Zipper
Leather	Red Polar fleece	Green cotton print	Diet coke can

Pig Skin – "stand-in" substrate to represent human skin due to similar fat, water, and hemoglobin content

Photo of samples

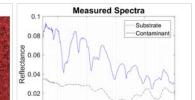


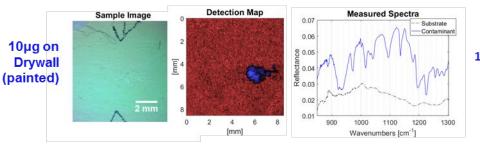


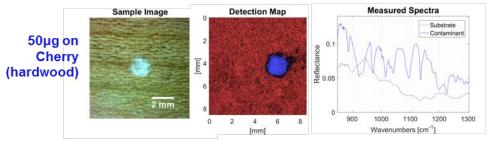
Representative Detection Examples

Aspirin on Various Substrates

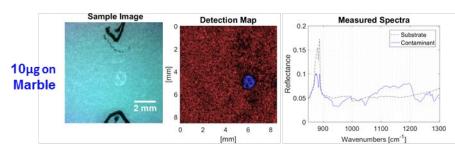
Sample Image Measured Spectra **Detection Map** Substrate Contaminant 10µg on Reflectar 90.0 40.0 Coke Can 0.02 2 mm Wavenumbers [cm⁻¹]

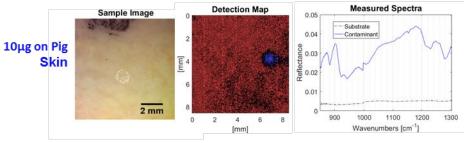


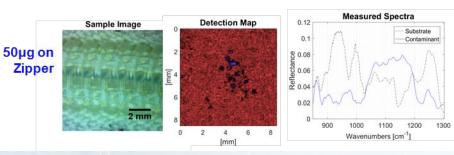




RDX on Various Substrates

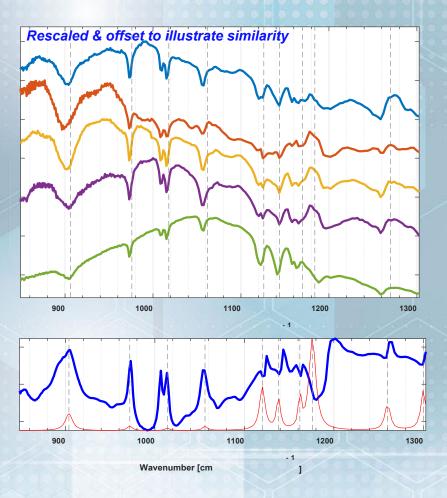








10μg Saccharin on Various Substrates

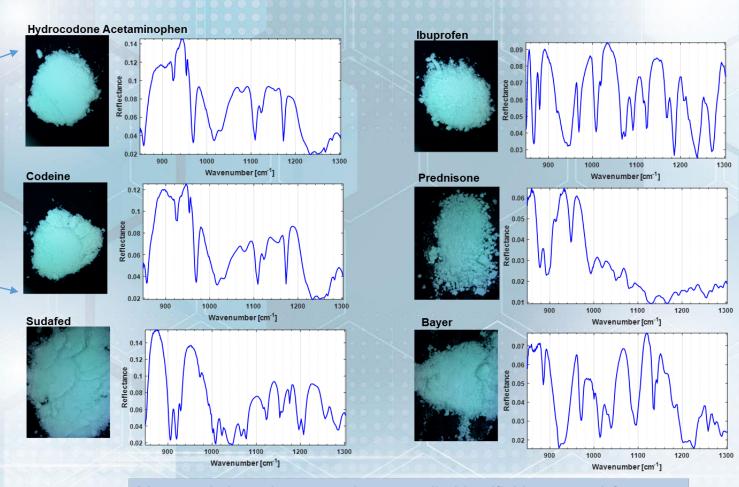


- on Leather
- on Cotton Cloth
- on Cherry Wood
- on Rubber Tire
- on Glass Tile
- Bulk Saccharin Powder (-In R)
- Saccharin absorption coefficient
 - Contaminant spectra is readily identifiable on each substrate
 - Features unique to saccharin are prominent despite spectral shape of underlying substrate



Bulk Powder Measurements of Drugs

Same drug, generic vs. name brand, shows spectra features of binder vs. active ingredient



Measured drugs have prominent, easily identifiable spectral features



Drugs in Plastic Bags

Bulk Powder in Mini-Bag

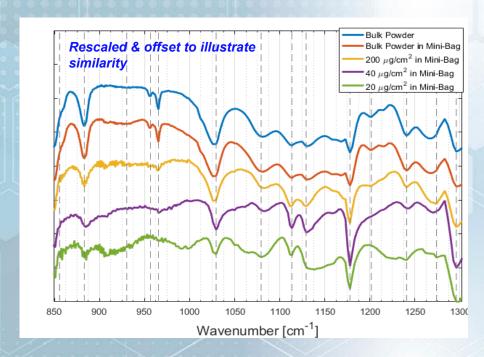


20 μg/cm² in Mini-Bag



Microscope closeup of 20 μg/cm² in Mini-Bag





- Successfully identified Sudafed (meth precursor) and benzocaine (fentanyl simulant) through a variety of plastic bags including clear Ziplock bags, Hefty black plastic trash bags, and thick clear plastic mini-bags
- Able to detect trace levels down to <20 μg/cm²
 - Roughly equivalent to an LOD of ~2 ng/pixel

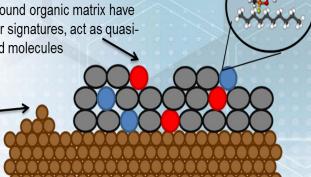


MORGOTH'S CROWN



Trace target residues in a background organic matrix have sharper signatures, act as quasiisolated molecules

Bare solid surfaces act as Lambertian scatterers & have broad, featureless absorption spectra



- Modeling of Reflectance Given Only Transmission or High-concentration Spectra for Chemical Recognition Over Widely-varying e**N**vironments
- Objective: Predict IR spectra of trace chemicals, taking into account the effects of target chemical loading and target chemicalsubstrate interaction
 - Major hurdle for active/passive standoff detection is spectral changes due to interactions with the substrate
 - Substrate/chemical combinations treated as separate library entries, leading to large libraries that can slow down detection algorithms
- Crowdsourced challenge to encourage breakthroughs in IR spectral modeling to predict trace spectra on a surface from bulk reflectance spectra



Spectral Prediction

Where we are now

- Most fielded libraries built from measurements & calibration targets
- Physics-based models work well for FITTING spectra, much less accurate for PREDICTING, too computationally intensive

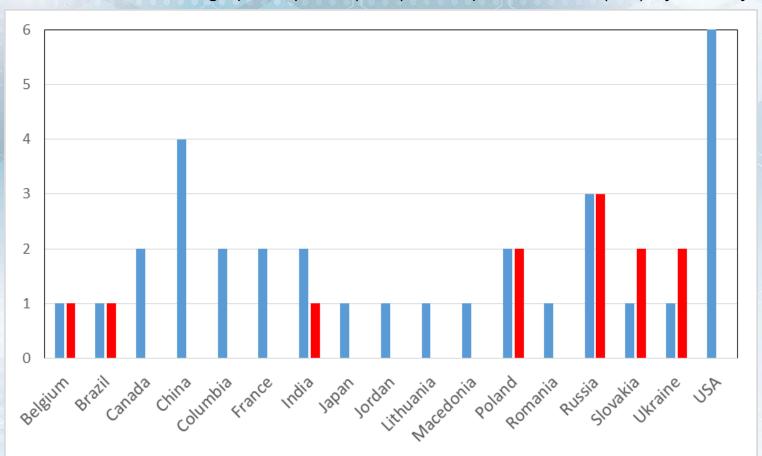
Where we want to go

- Accurate prediction of spectra at any loading, particle size, or morphology based on measured substrate & bulk spectra
- Tie-in to "physics" desirable, but not the "end all" first priority is an algorithm that WORKS, and is computationally efficient
- What can the physical chemistry community learn from the machine learning / statistics / computer science community?



Geographical Challenge Statistics

Distribution of challenge participants (blue) and top ten scorers (red) by country





Top Scorer Data

prize _ winners

Solver "Handle"	Country	Provisional Score	Final Score	Topcoder Ranking	Approach
Ap31	Russia	815,203.28	716,755.76		Random Forest
ZFTurbo	Russia	759,454.94	710,956.29		Average of Closest Matching Spectra
Albanator30	Belgium	773,226.29	701,113.19		Target/Substrate Ratio
cannab	Russia	800,475.55	698,732.62		Gradient Boost Decision Trees
05_Tomek.Dyczek	Poland	792,879.36	695,729.51		Extra-Trees Regressor
Wleite	Brazil	749,591.58	679,832.19		Random Forest
Brev	Ukraine	793,919.02	672,733.49		Random Forest
Nickil21	India	778,851.06	668,196.28		Extra-Trees Regressor
Nofto	Slovakia	801,048.34	662,618.02		Random Forest
Mloody2000	Poland	708,798.28	557,166.54		Random Forest

То	pcoder Rating
	2200+
0	1500-2199
	1200-1499
	900-1199
0	< 899

(U) Small training set compared to test set meant methods used had to be robust against overfitting. Two of the winning algorithms did not use machine learning, simple averages proved more accurate.



MC Challenge Take-Home Points

"Physics agnostic" spectral prediction DOES work, though there are limits to how far a given model can be extended

- Better results with substrates that have "weaker" intrinsic features
- Better results with "more similar" chemicals (functional groups, location of spectral features)
- Hybrid algorithms that use physics-based constraints on training likely the most efficient

Machine learning approaches need more data

- Better binning of metadata into separate training classes
- Figure out which parameters are easily predictable, and which are best to train for



Next Steps



Program Next Steps

- October 2018 test against chemical and biological simulants (aerosol and surface deposition) at Dugway Proving Grounds
- November 2018 Phase 2, 2nd benchmark test at JHU/APL (indoor trace surface detection at 5-25 meters)
- March 2019 Phase 2, 3rd benchmark test at JHU/APL (outdoor trace surface detection at 5-25 meters)
- May 2019 Field test screening backpacks & vehicles for explosives other hazards at Indy500







Questions?