



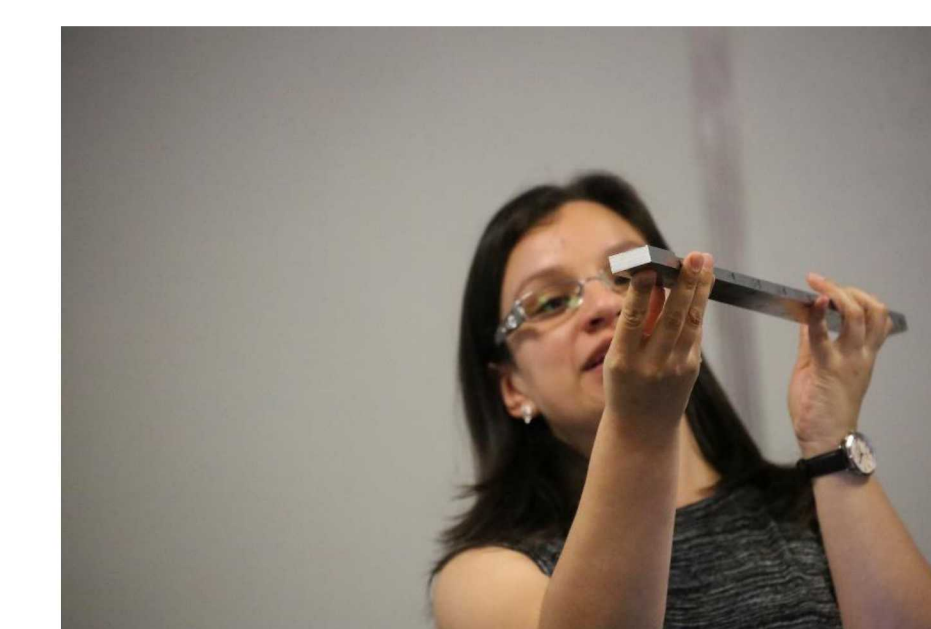
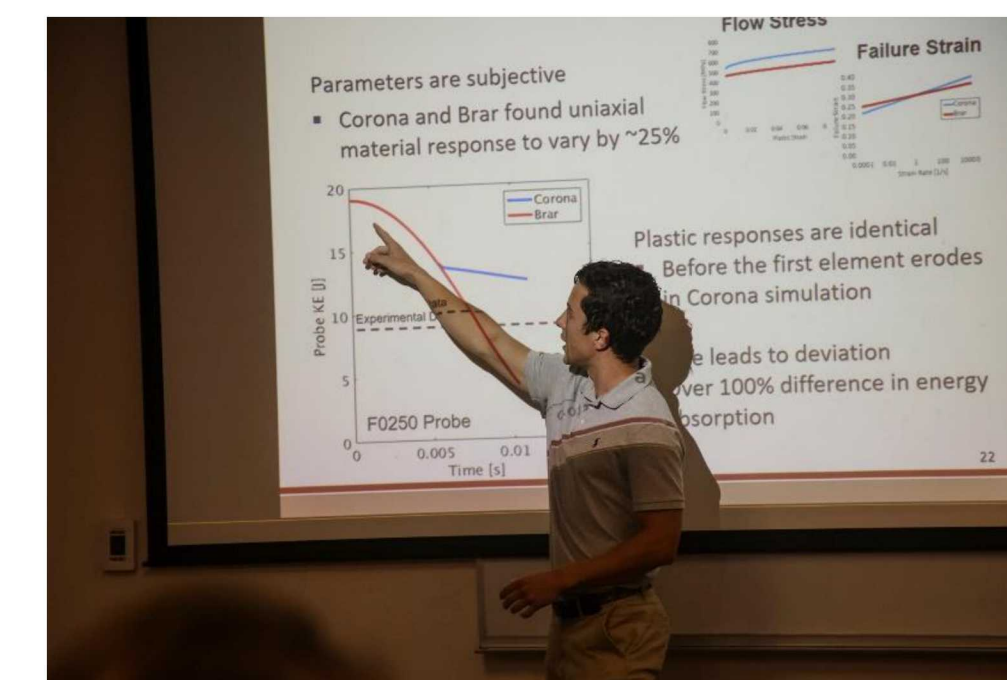
## Overview / History

- Founded in 2014, NOMAD is a collaborative and educational research institute that unites graduate and undergraduate students to work on challenging research problems in engineering sciences
- Institute is co-hosted by Sandia National Laboratories and the University of New Mexico
- Inaugural year (2014) held at Sandia National Laboratories; 2015 and beyond held at the University of New Mexico Campus
- On average, each year has 6 projects consisting of 3 students and 2-4 mentors



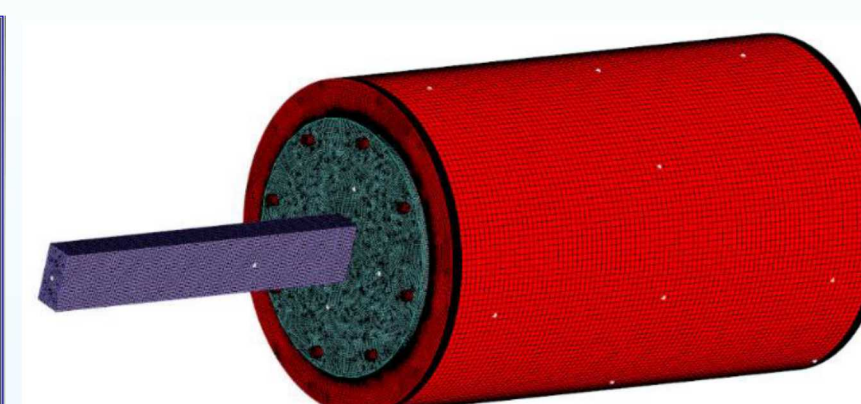
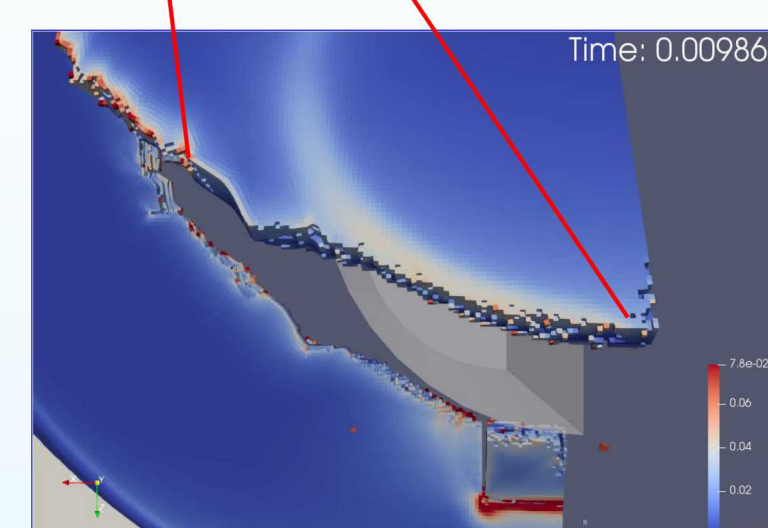
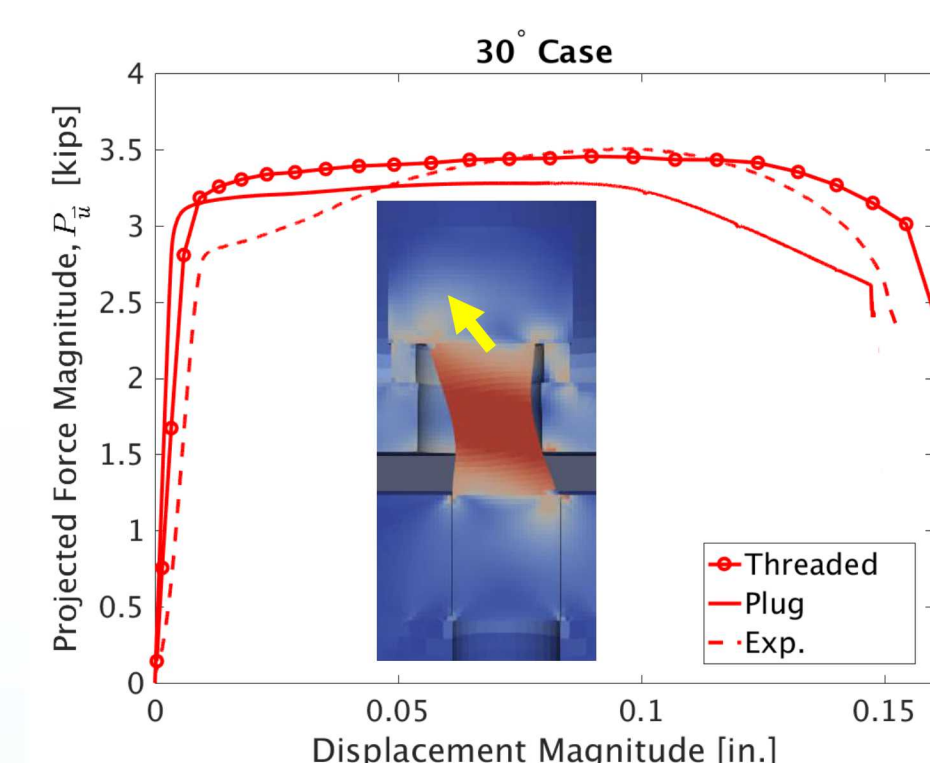
## 2018 Highlights

- Research projects possessing experimental and computational aspects
- Weekly technical seminars on topics related to nonlinear mechanics and dynamics from visiting professors
- Organized social events including a night at an Albuquerque Isotopes baseball game
- Students presented research discoveries at the final NOMAD technical seminar



## Impact

- Publish and present research in challenging fields of nonlinear mechanics and dynamics
- Ability to network and collaborate with researchers from around the world
- Work on cutting edge research performed at a national laboratory



## Path Forward

- Continue to improve external collaborations with greater research community by solving challenging research problems
- Develop strategic pipeline of highly qualified candidates to support national security mission
- Stewardship of novel technical advances to improve understanding of complex mechanical systems under extreme environmental loads

## Contact Information

Visit our website at:

[http://www.sandia.gov/careers/students\\_postdocs/internships/institutes/nomad.html](http://www.sandia.gov/careers/students_postdocs/internships/institutes/nomad.html)

Contact the organizers at:

[nomad@sandia.gov](mailto:nomad@sandia.gov)

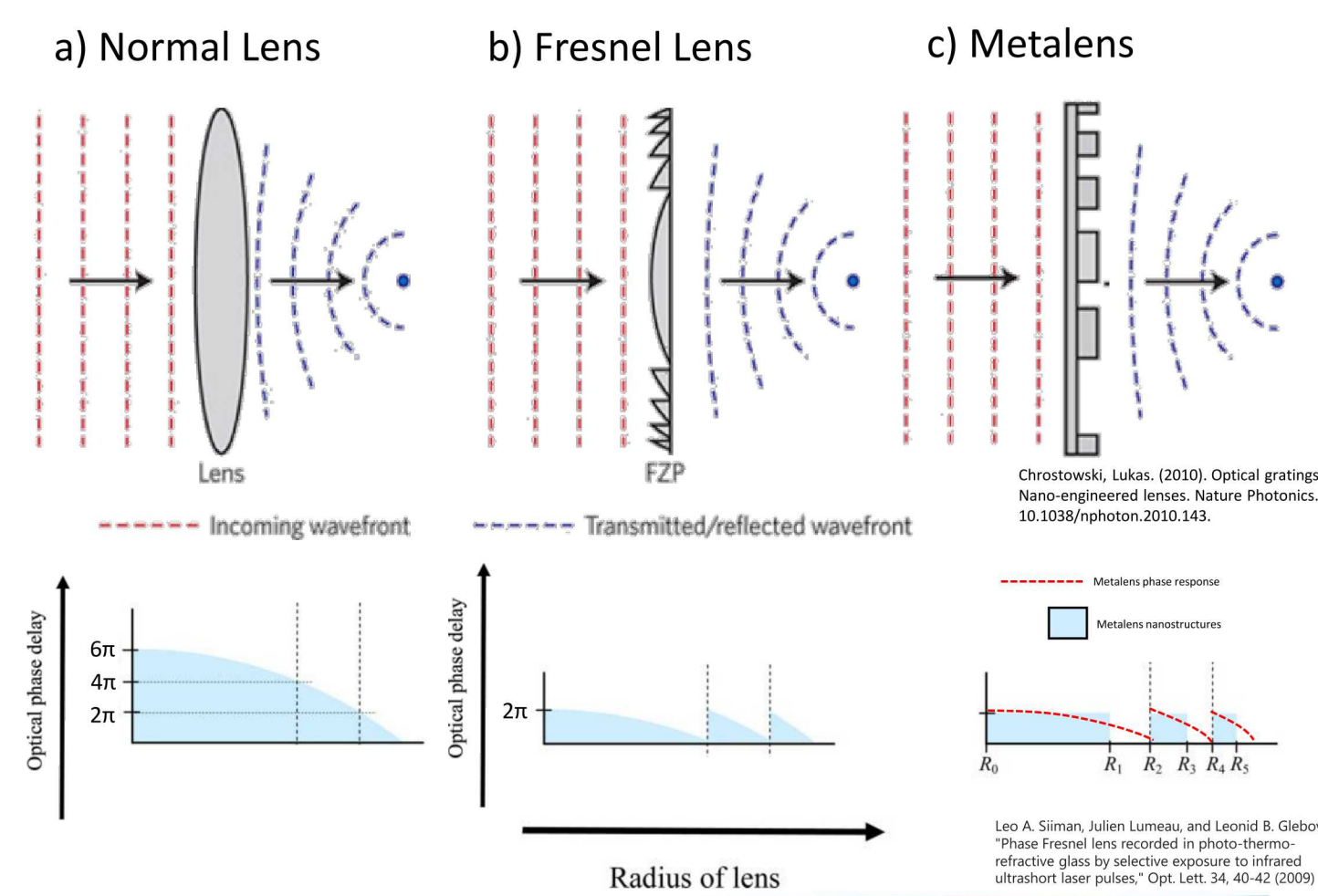




## Introduction / Motivation

### What is a Metalens?

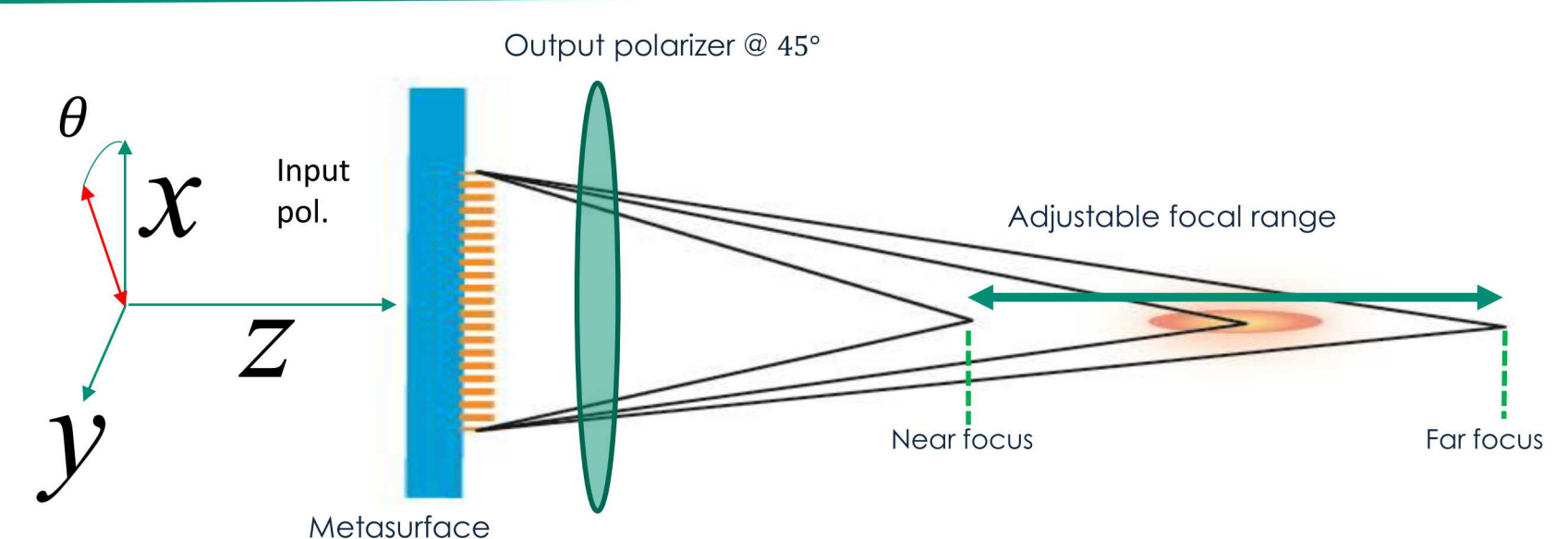
- A normal optical lens can be thought of as a parabolic phase manipulation of incident light.
- A Fresnel lens creates the same effect with less material by keeping the same slope of this parabolic phase and limiting the overall phase change to  $2\pi$ .
- A Metasurface lens, i.e. Metalens, is a sub-micron thick, nanofabricated device that can mimic a lens and include additional functionality.



### Goal of Project:

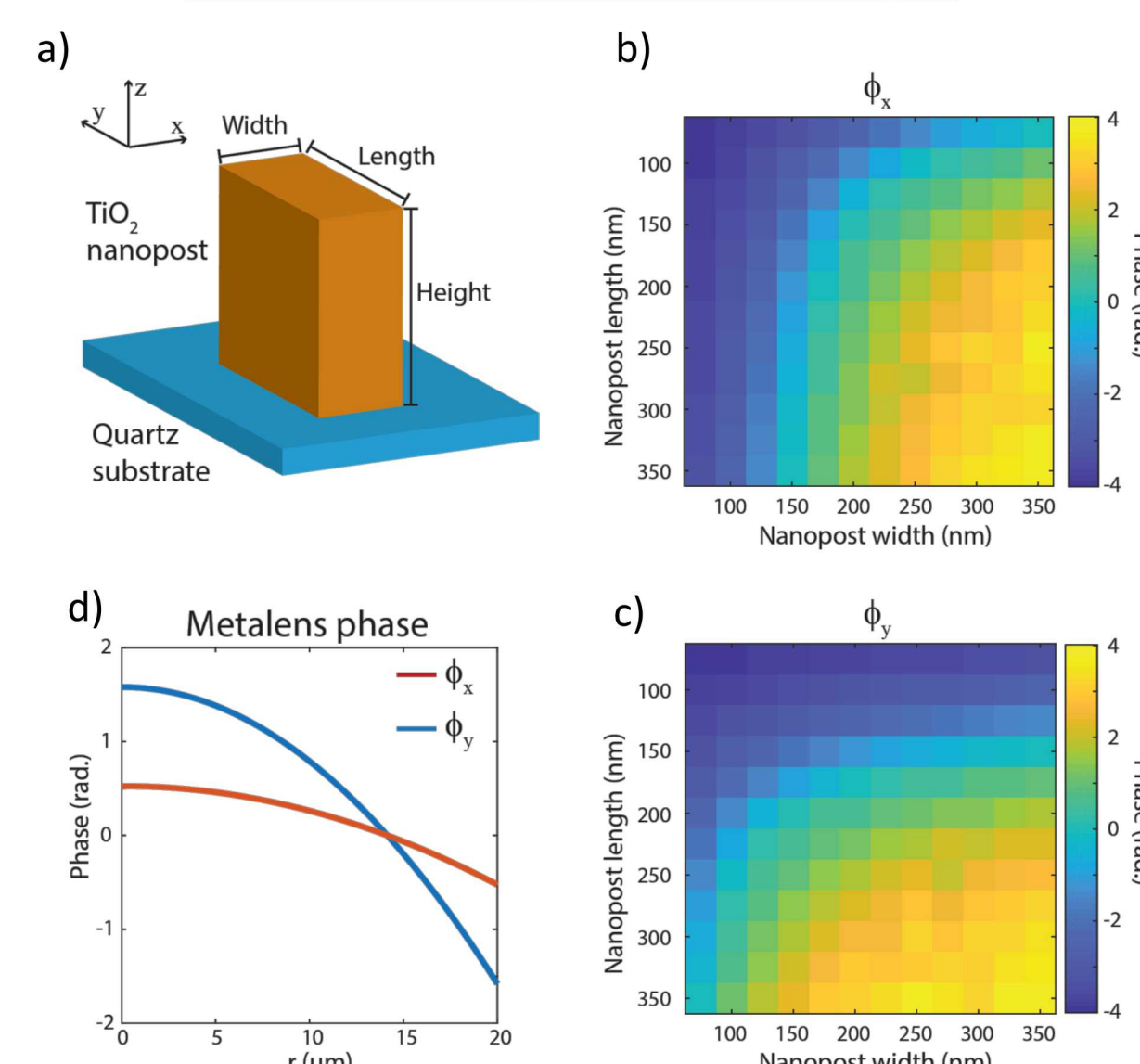
*Demonstrate varifocal, Achromatic Metalenses for imaging/display*

- Vary the linear input polarization from  $0^\circ$  to  $180^\circ$  to vary the focal length of the Metalens



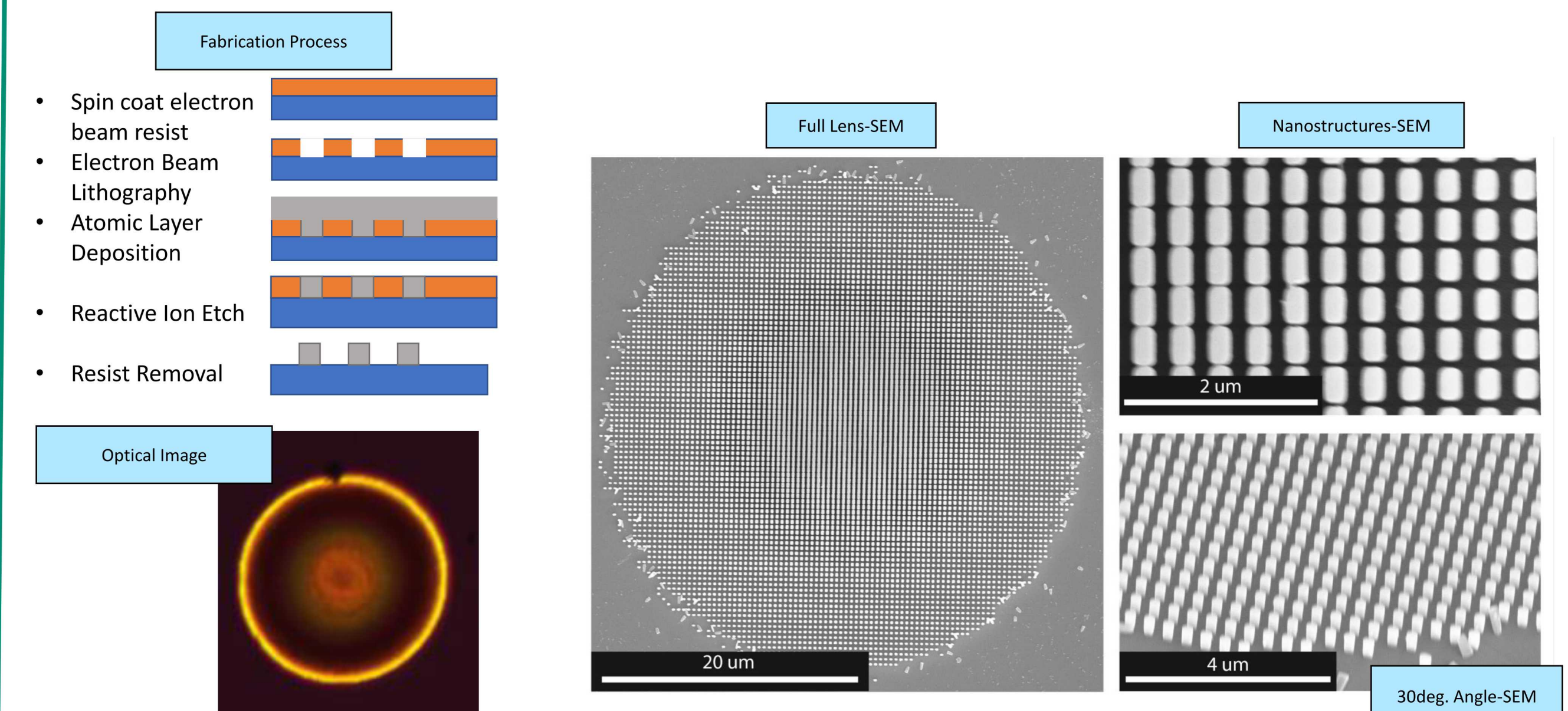
## Approach

### Electromagnetic Design



- Phase response of different nanopost dimensions is determined using FDTD simulation.
- Asymmetric nanopost dimensions lead to a different phase response from x-polarized light vs. y-polarized light.
- We map phase information to two different parabolas, one for x/y-polarizations, resulting in a continuously-tunable polarization-dependent focal length.

### Nanofabrication



**Fabrication Process:** Carried out at the SNL Center for Integrated NanoTechnologies (CINT).  
**Optical Image:** Color microscope image of Metalens acquired using experimental setup at UNM's Center for High Technology Materials (CHTM).  
**SEM Images:** (Left) Top-down image of the full Metalens. (Upper right) Magnified image shows asymmetric, polarization-sensitive nanostructures. (Bottom right) 30deg. projection SEM used for nanostructure height measurements – estimated height ~ 600 nm.

## Challenges

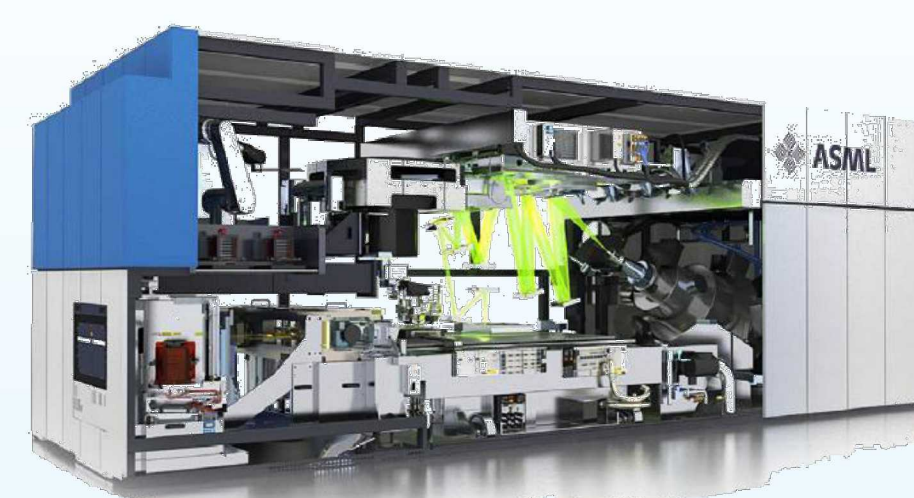
### Scaling

- Electron beam lithography does not scale well to larger diameter optics.
- Need to transition to a mask-based photolithography scheme for rapid fabrication.



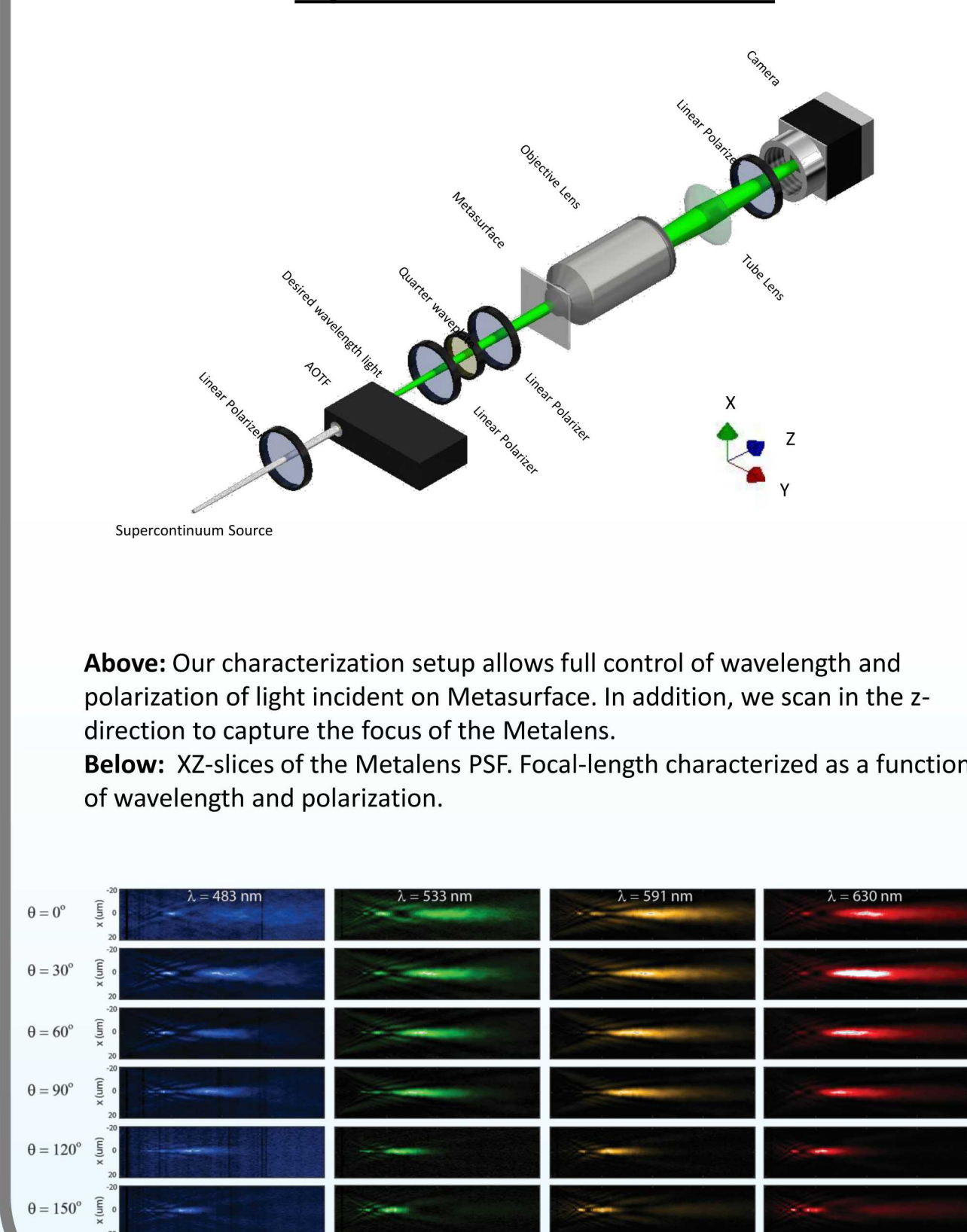
Electron Beam Lithography

Deep UV Lithography

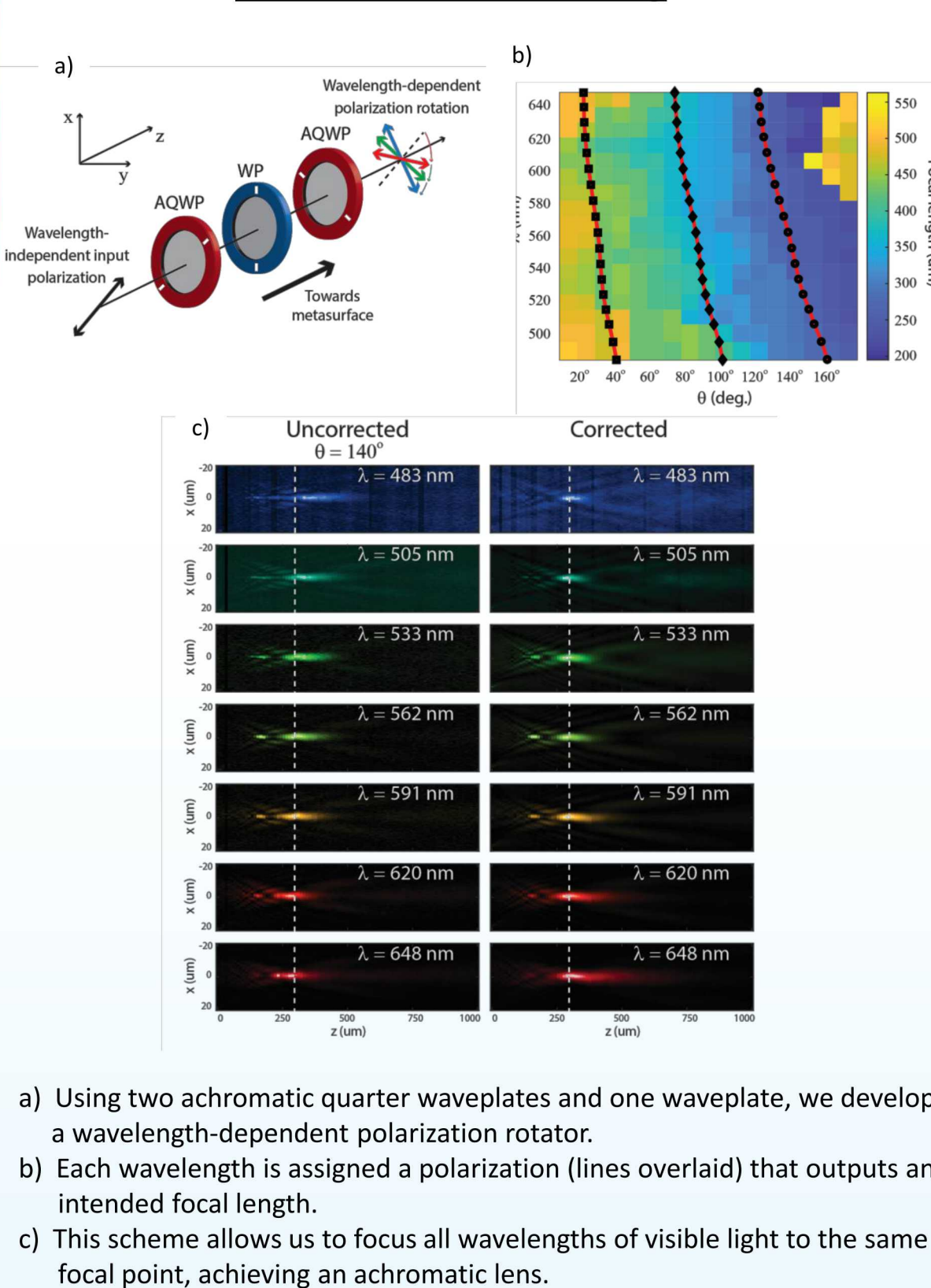


## Results

### Optical Characterization

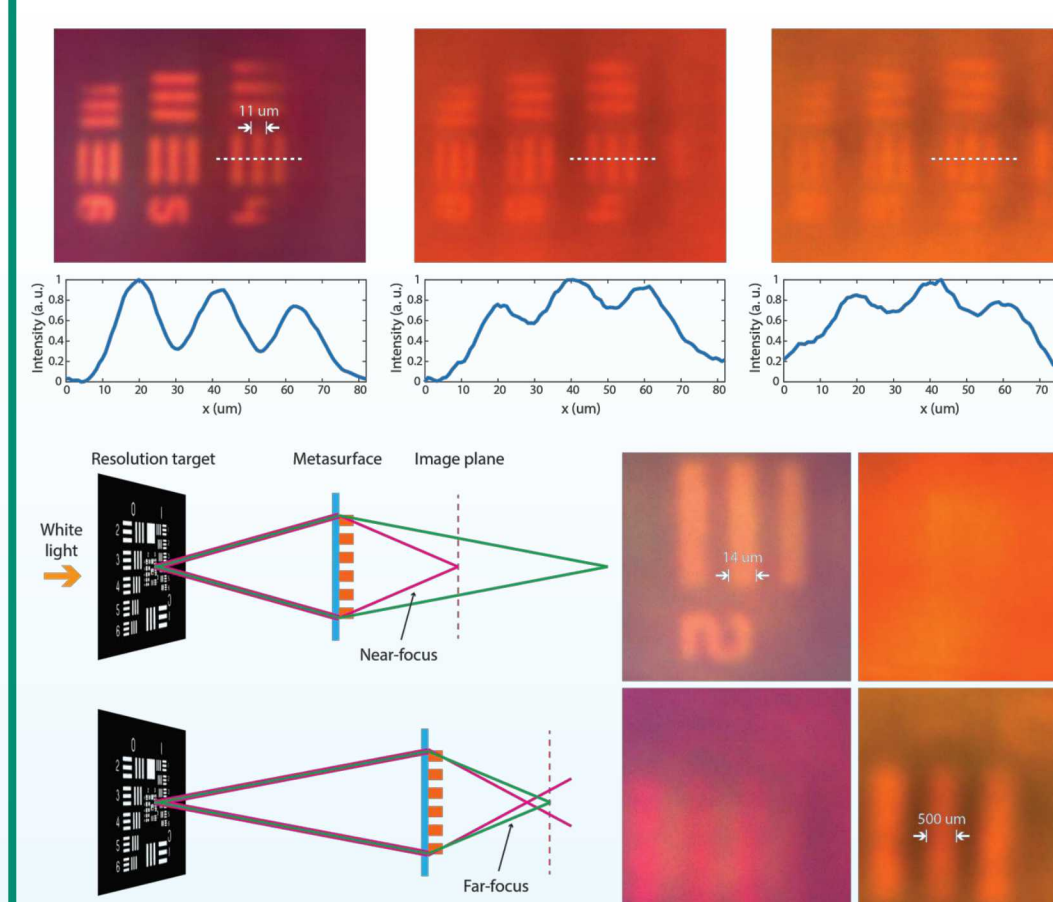


### Achromatic Focusing



### Broadband Imaging

- Image-capture using white light proves the effectiveness of our Metalens.
- Imaging a standard 1951 USAF resolution target we can see:
  - top:** the resolution as a function of different focal settings.
  - bottom:** Polarization-based refocusing.



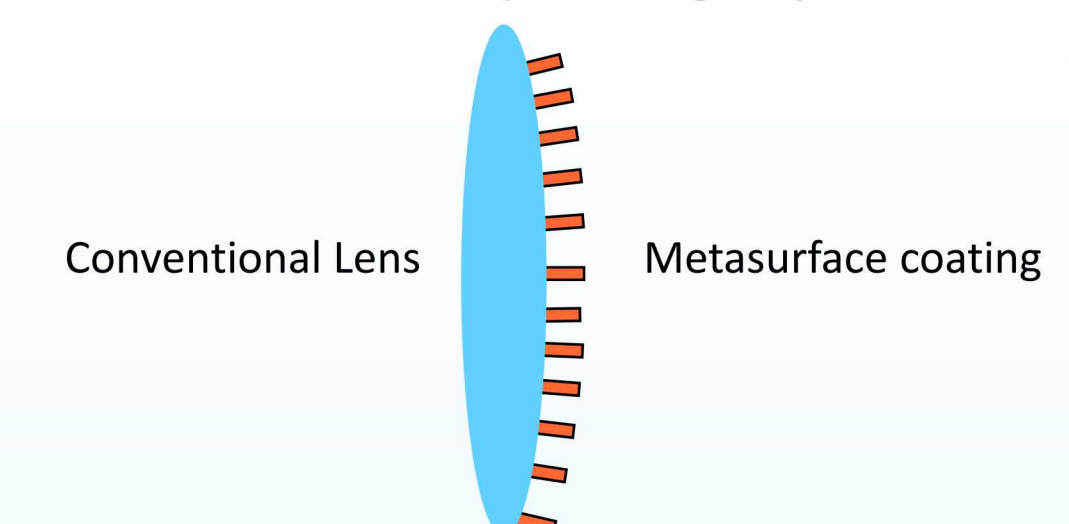
## Partnerships

### Collaborations

- This work was completed in collaboration with Google Daydream for a Virtual/ Augmented Reality display application.
- The work was completed as a joint UNM and Sandia National Laboratories effort. Adam Backer acknowledges research support from the Harry S. Truman Fellowship.

### Future Work

- Patterning Metasurfaces onto conventional optics to further increase the functionality of a single optical device.
- Use computational inverse-design techniques to develop non-intuitive optimized metasurfaces.





# Sandia National Laboratories Irradiation Resistance of Interfaces in Zr–Nb Nanocomposites

Elton Chen<sup>[1,2]</sup>, Rémi Dingreville<sup>[1]</sup>, Chaitanya Deo<sup>[2]</sup>

[1] Center for Integrated Nanotechnologies, Sandia National Laboratories

[2] G.W. Woodruff School of Mechanical Engineering, NRE Program, Georgia Institute of Technology



## Introduction / Motivation

- Grain and phase boundaries act as defect sinks and barriers, subsequently strengthen the material by governing different deformation mechanisms like interfacial shearing, dislocation nucleation and motion, and interfacial-driven twinning.
- Nano-laminate composites are engineered to take advantage of these boundary effects and further extend material mechanical performance.
- Exposure to irradiation and other external factors can change phase boundary structures, compositions, and properties.
- Mechanical performance degradation can be tied to microstructural change at/near the phase boundaries.

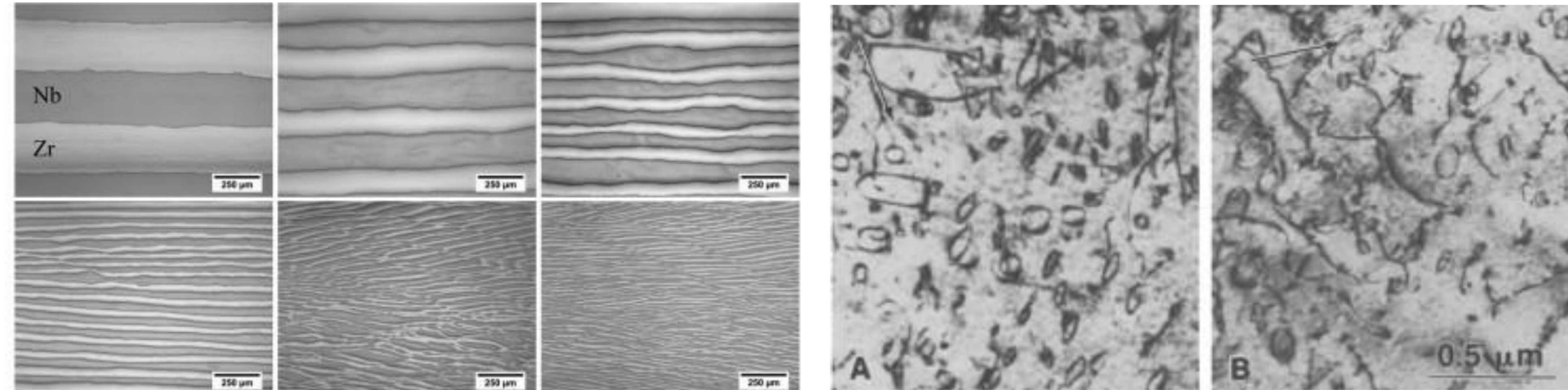


Figure 1: Rolling manufactured Zr-Nb laminar composites. [1]

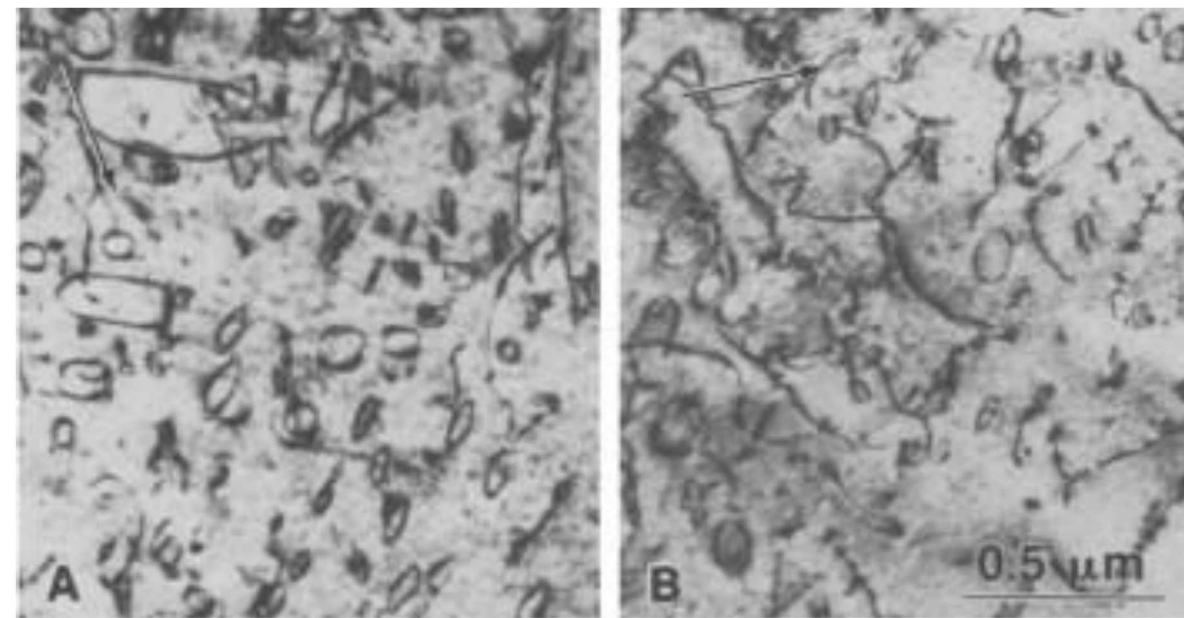


Figure 2: Irradiated dislocation loop in bulk Zr. [2]

## Modeling Approach

- Molecular Dynamics (MD) is the best method of simulating irradiation defect accumulation at the atomic size-scale.
- Each radiation event is represented by a series of atom-atom collision events combined into a larger displacement cascade.
- Simulation of full displacement cascades is too slow and computationally expensive for high dose level.
- Permanent radiation damage are directly introduced as vacancy-interstitial (Frenkel) pairs by atomic displacement. [3]
- Accelerated defect implantations allow access to high dose damaged states.
- Simulation is relaxed intermittently to allow for defect structure evolutions.
- Initial nano-laminate composites are constructed by rotating and stacking Zr and Nb bulk crystals with structured interfaces.

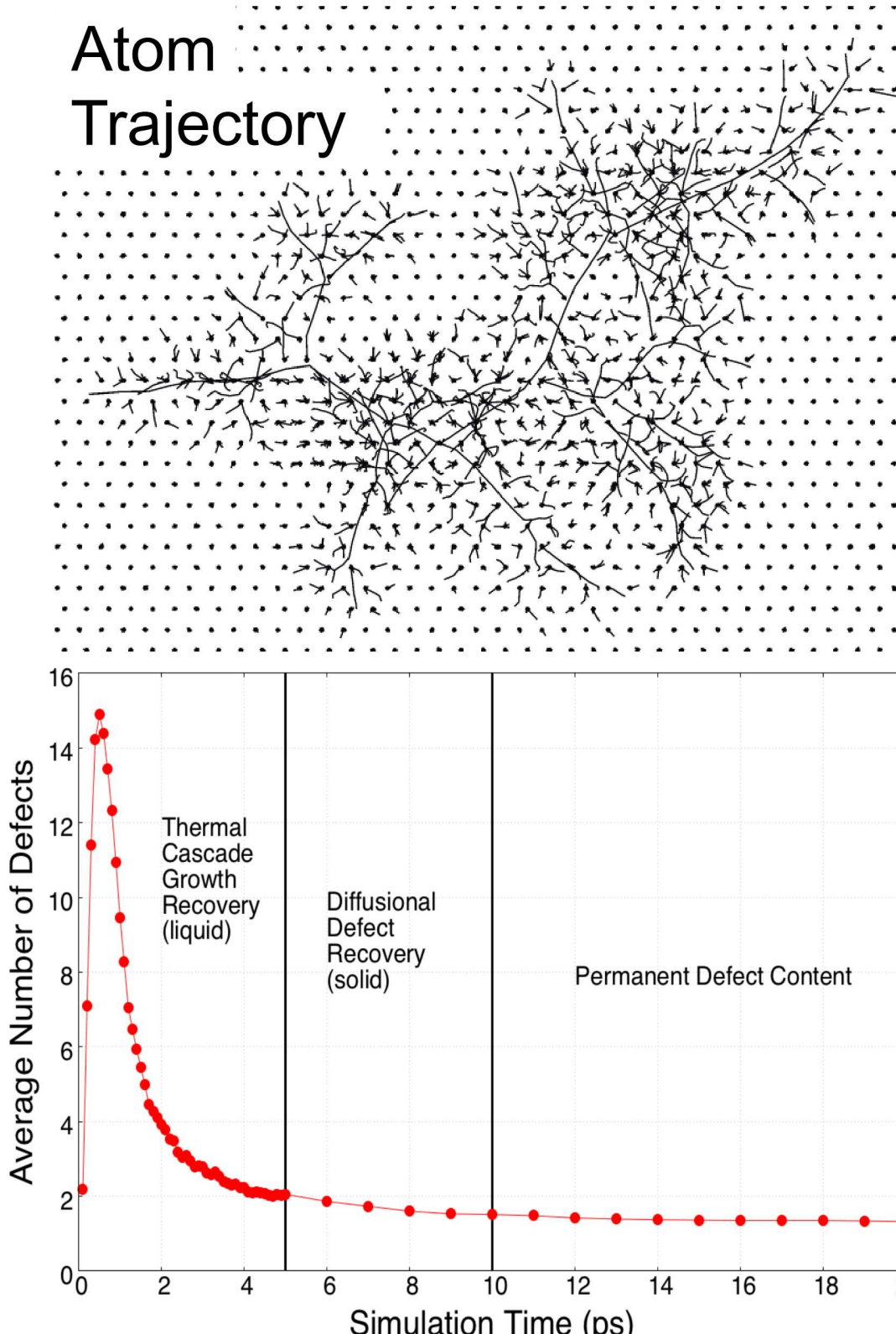


Figure 3: Radiation cascade development & defect population.

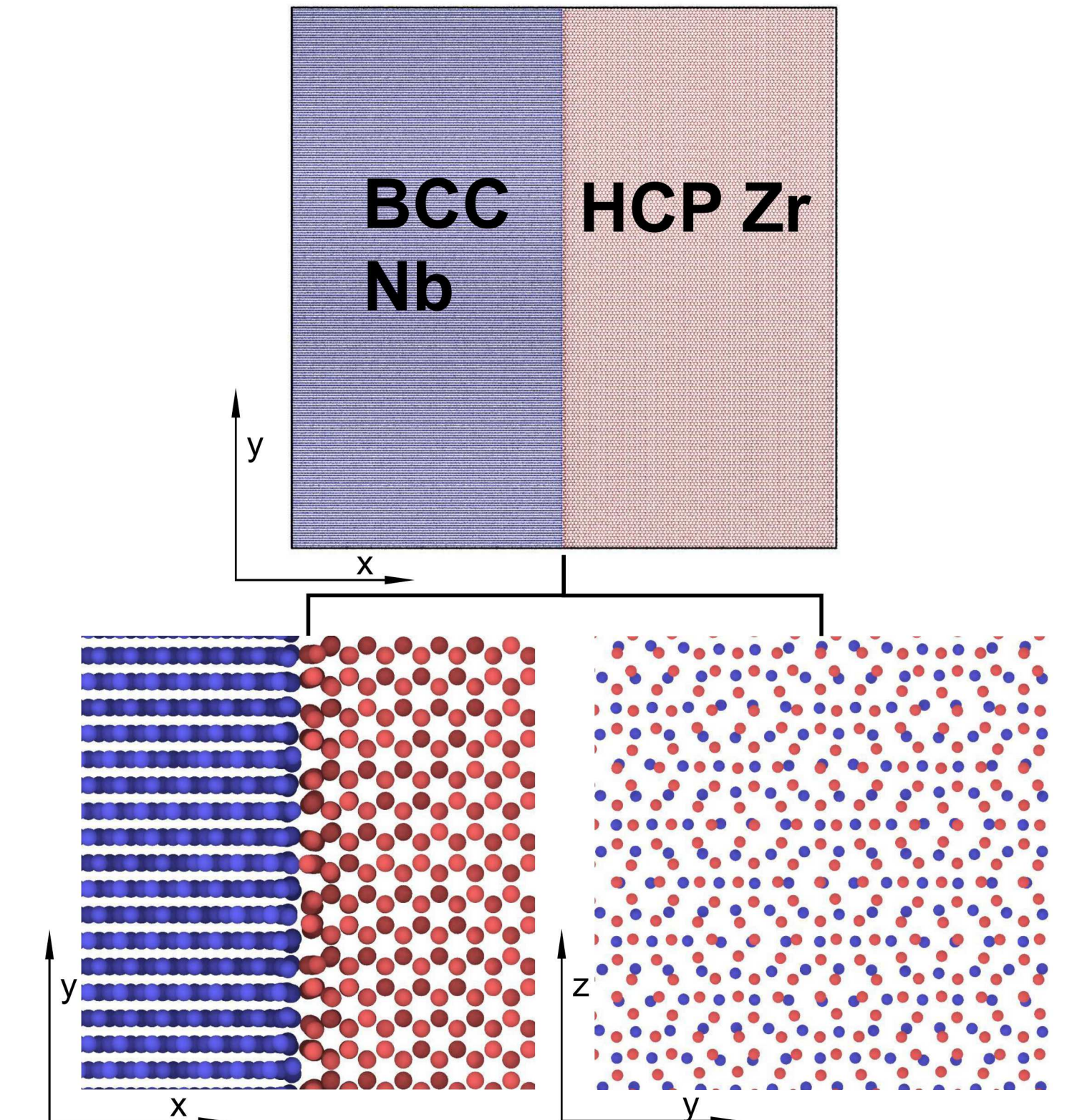


Figure 4: Zr-Nb laminate composite & interfacial microstructure.

## Results

### Bulk Crystalline

- Three stages of defect accumulation:
  - Small dislocation accumulation
  - Dislocation saturation
  - Coalescence to large dislocations
- Formation of larger dislocation loops decreases the overall dislocation density.
- HCP Zr dislocation formations match those observed from electron radiation experiments.

### Nano-laminate Composite

- A secondary rise in dislocation density exists in the BCC sub-layer.
- Increase in dislocation density corresponds to trapping of smaller BCC dislocation loops near the phase boundaries.
- Composition mixing near the boundary induces HCP → BCC phase transition and dislocation trapping.

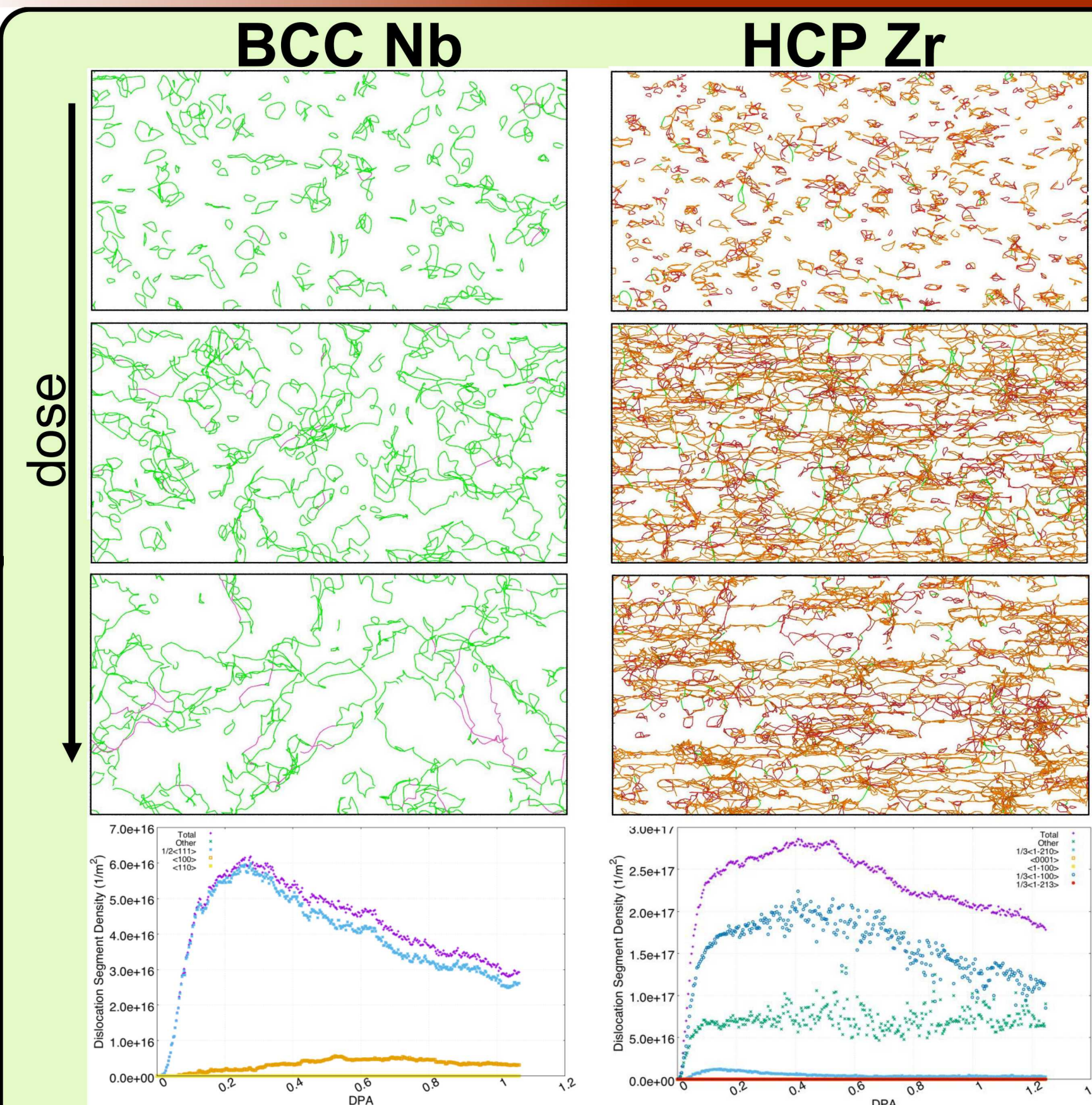


Figure 5: Dislocation accumulation & evolution in bulk Niobium and Zirconium.

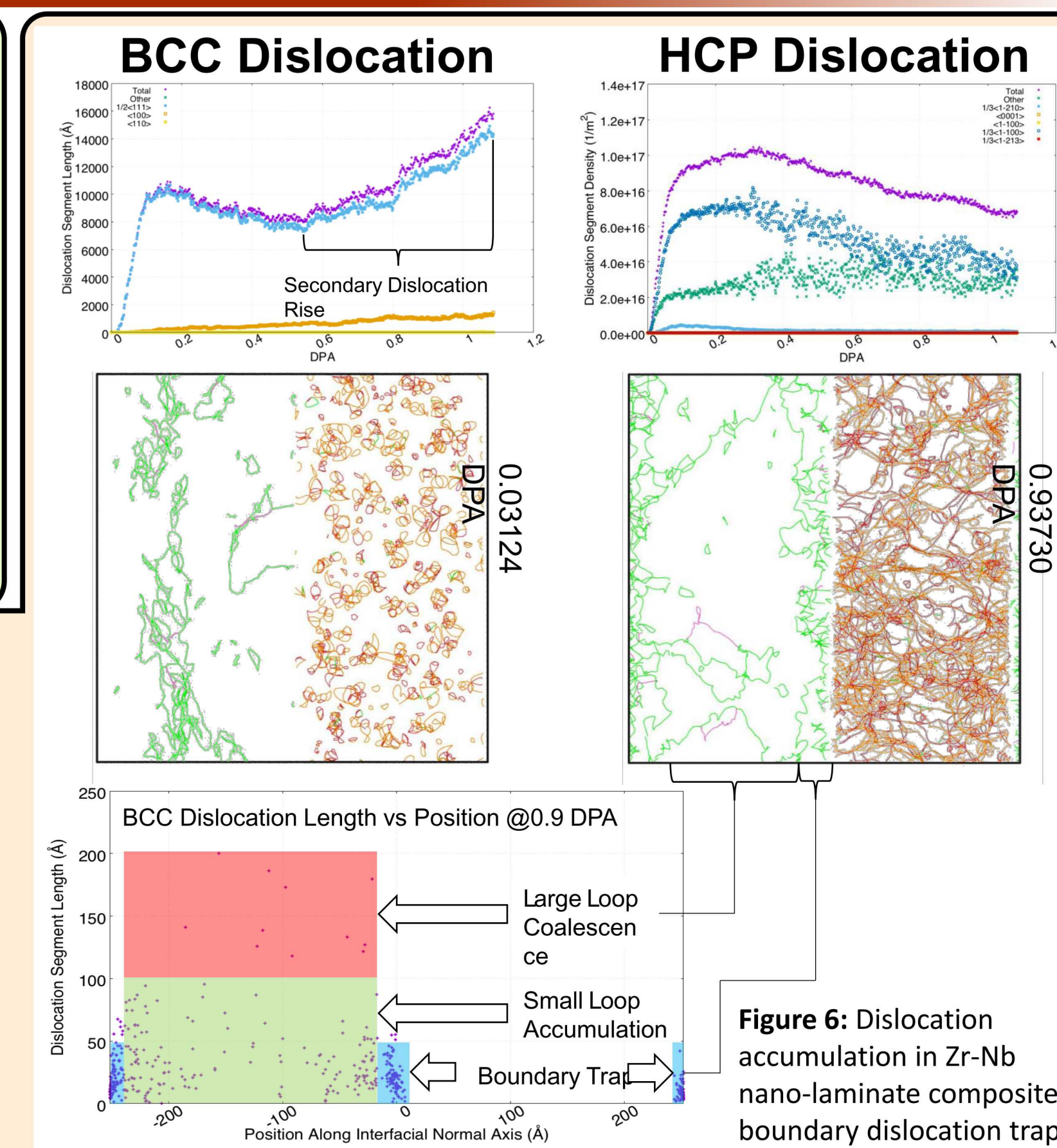


Figure 6: Dislocation accumulation in Zr-Nb nano-laminate composite & boundary dislocation trap.

## Partnerships

- Improving the radiation damage accumulation model by introducing new dose rate and dose type effects.
  - Evaluating the radiation effect on mechanical performance through fracture simulations of the irradiated nano-laminate composites.
- This work was performed in part at the Center for Integrated Nanotechnologies, an Office of Science user facility operated for the U.S. Department of Energy. The authors acknowledge the support from the Woodruff Faculty Fellowship at Georgia Institute of Technology, and the Sandia Laboratory Directed Research and Development (LDRD) Academic Alliance program.

- [1] Knezevic, Marko, et al. "Texture evolution in two-phase Zr/Nb lamellar composites during accumulative roll bonding." *International Journal of Plasticity* 57 (2014): 16-28.
- [2] Choi, Sang Il, and Ji Hyun Kim. "Radiation-induced dislocation and growth behavior of zirconium and zirconium alloys—a review." *Nuclear Engineering and Technology* 45.3 (2013): 385-392.
- [3] Chartier, A., et al. "Early stages of irradiation induced dislocations in uranium." *Applied Physics Letters* 109.18 (2016): 181902.





## Introduction / Motivation

Physical vapor deposition (PVD) is used to produce explosives with great control over the morphology and microstructure. Deposition of densified PETN films is possible by depositing onto substrates with high surface energy. However, the densified films suffer from micro cracking throughout, as shown in Figure 1, because of residual thermal stresses caused by a mismatch of thermal expansion coefficients between the substrate and the PETN. To simulate the effects of the micro cracking, gaps are engineered by separating two films by a known width as shown in Figure 2. Refractive imaging is used to observe the shock front shape to determine if the detonation propagates across each gap thickness.

Figure 1. Densified PETN film displaying micro cracking

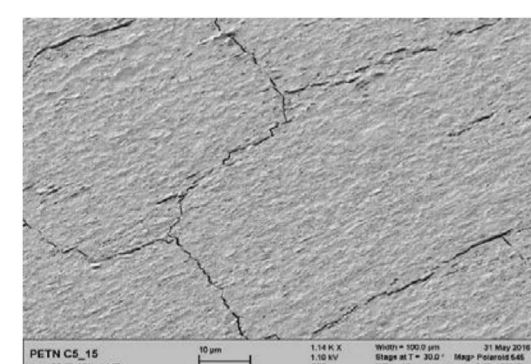
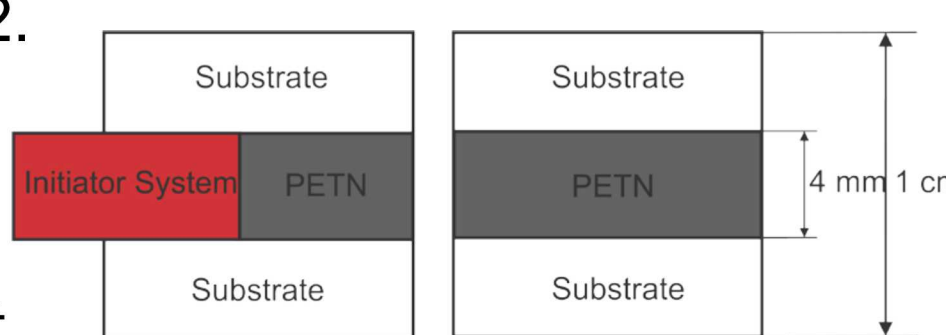


Figure 2. Top down view of gap test structure



## Approach

Refractive imaging was used to visualize shock waves in air produced by the detonation of the PETN films. The primary imaging method was a “focused” shadowgraph technique. Focused shadowgraph is used to visualize the second derivative of the refractive index. A detailed schematic of a focused shadowgraph system is shown in Figure 3.

The shock front was tracked using a custom developed image processing routine. The shock tracking was done with a Canny filter, and then a mask was applied to remove noise that was inadvertently picked up with the Canny filter. The approach ultimately resulted in an image of the shock front location. The shock front was automatically extracted from the binary image for each pixel row in the image. The pixel location was converted to a spatial location using a calibration grid with known spacing. Velocities are calculated from successive shock wave positions using a backward finite different method. Where appropriate, centered differences were used for steady state shock wave propagation.

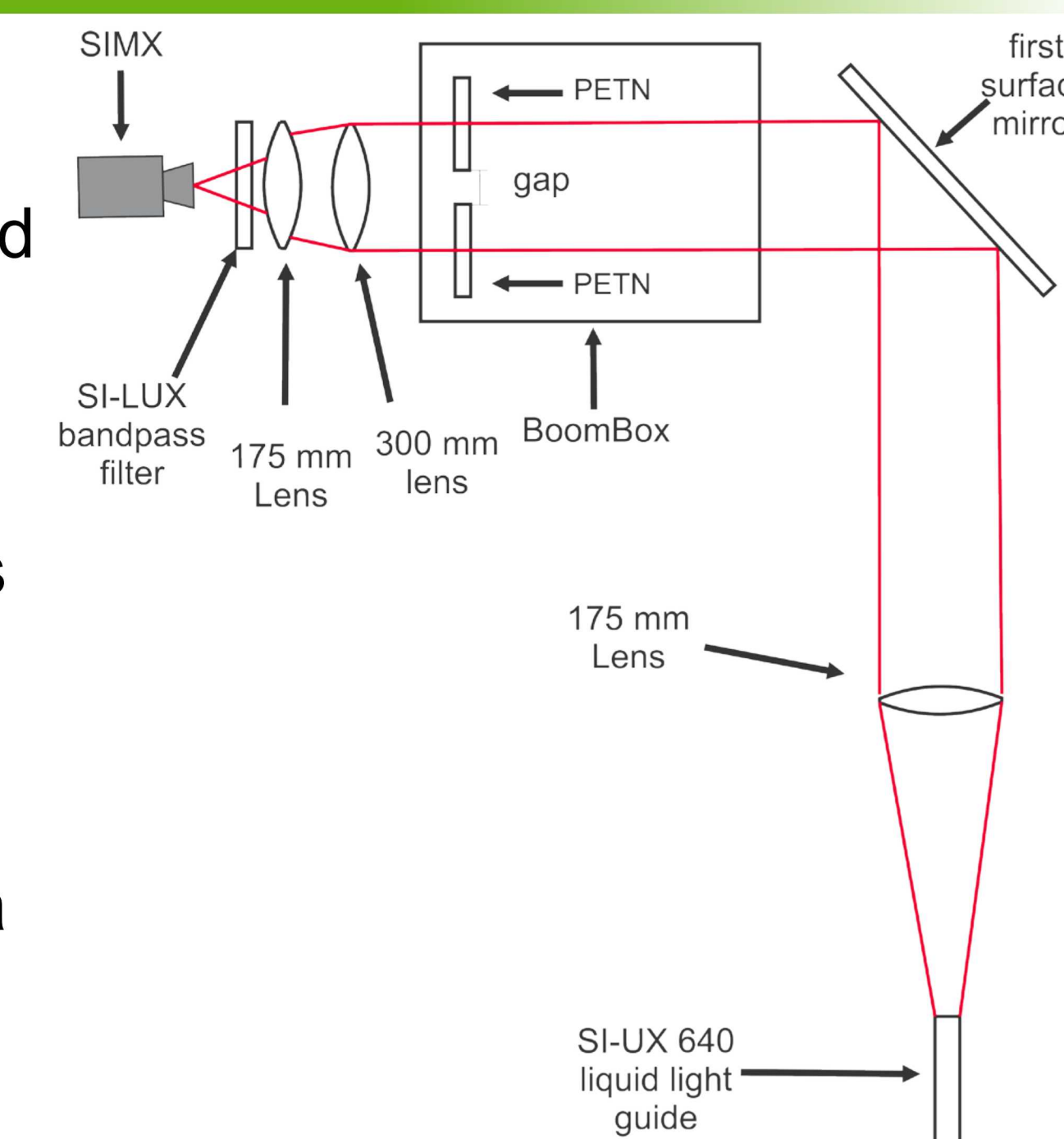


Figure 3. Focused shadowgraph system used during testing.

## Results

Through the test series, the critical gap width for reliable detonation of a 200  $\mu\text{m}$  thick film was found to be approximately 75  $\mu\text{m}$ . Detonation across a gap larger than 80  $\mu\text{m}$  was observed in one instance. The tests with successful propagation exhibit a deceleration of the shock front and then return to steady state as shown in Figure 4. When the gap is large enough detonation fails to propagate, and the shape of the shock front goes from roughly linear in shape to roughly circular in shape as shown in Figure 5.

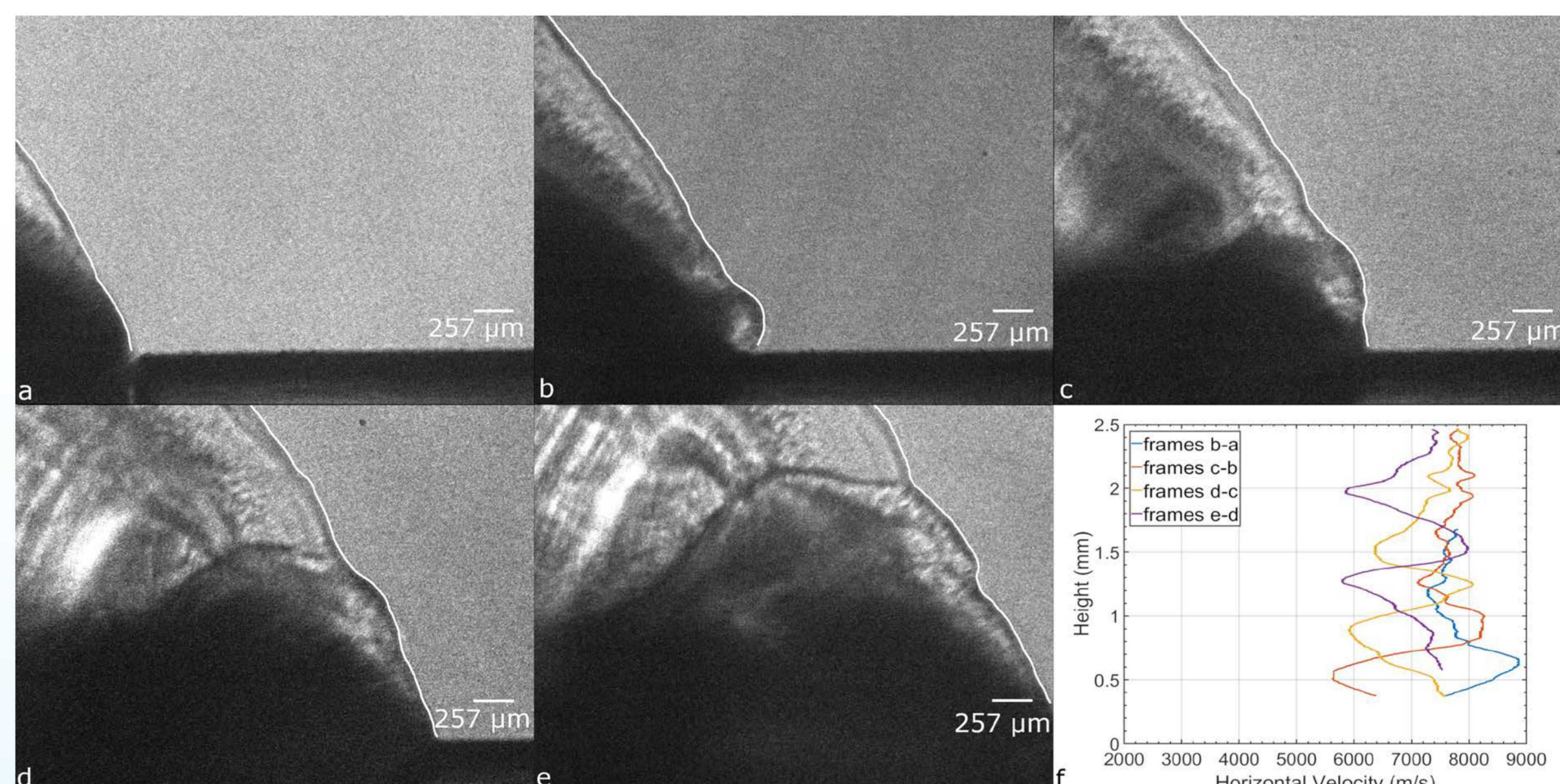


Figure 4. (a)-(e) Propagated shot, 90 ns between frames. (f) Backward difference horizontal velocity for frames above.

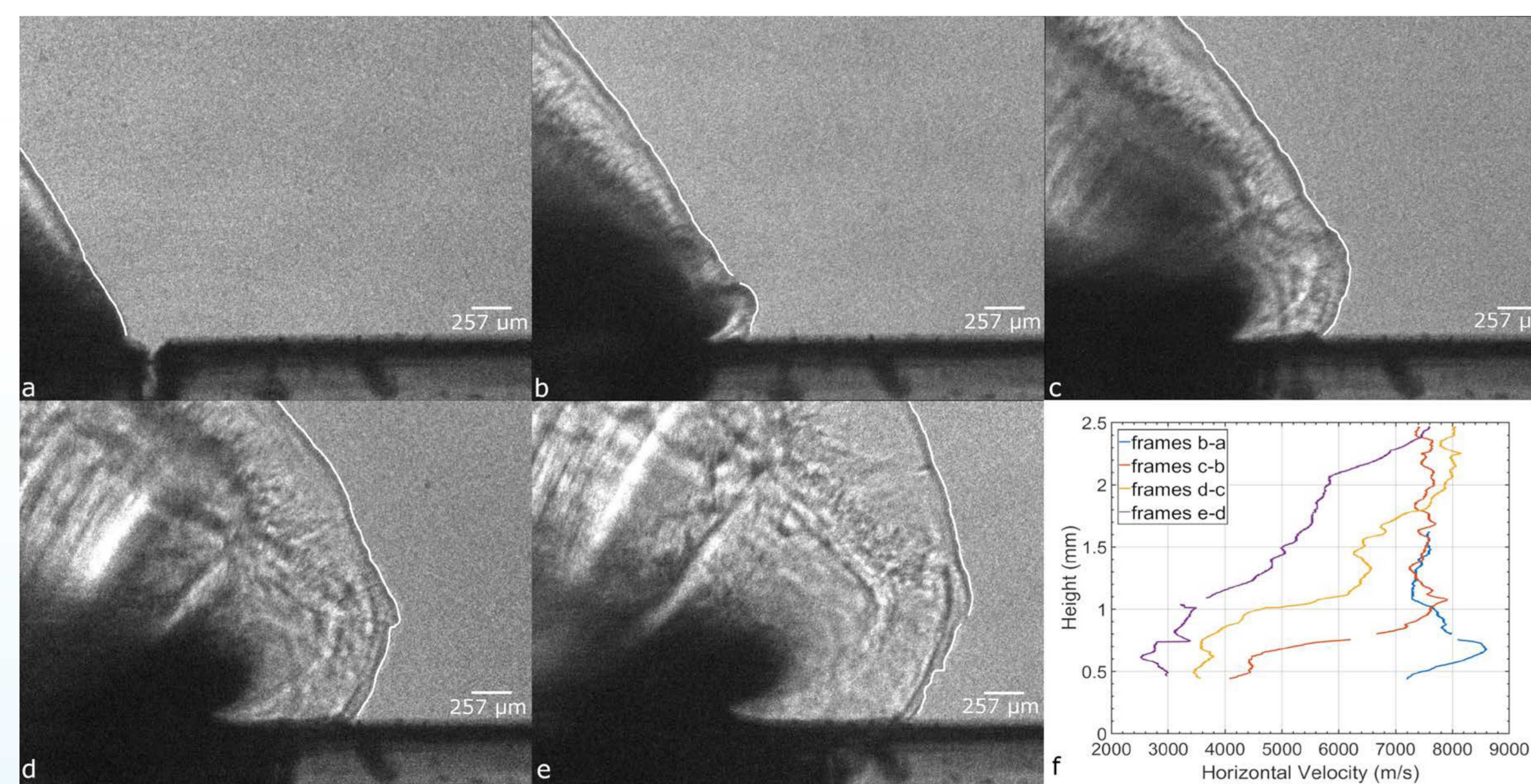


Figure 5. (a)-(e) Failed shot, 100 ns between frames. (f) Backward difference horizontal velocity for frames above.

## Future Work

Implementation of velocity measurements normal to the shock front is currently in development. This is achieved by matching a point on a shock front to the point in the normal direction on the subsequent shock front. An example is shown in Figure 6.

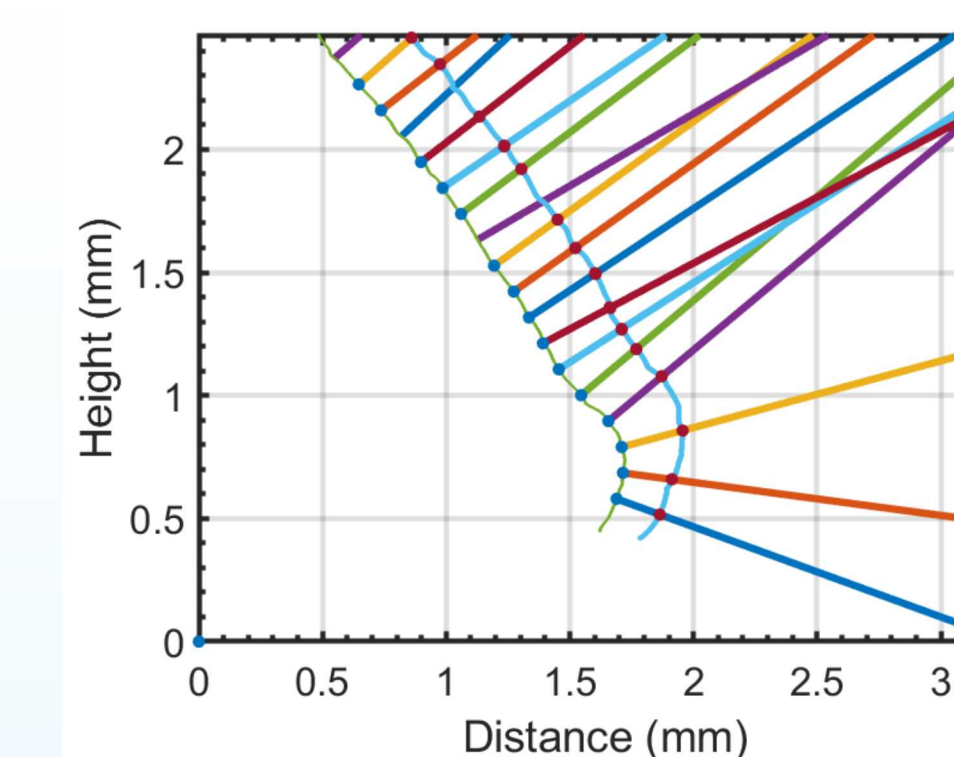


Figure 6. Current state of normal point matching. The line connects a point on the original shock front to two possible matches on the second shock front.



# Sandia National Laboratories Autonomy-Enabled, Real-Time, Rapid Trajectory Generation for Highly Dynamic Hypersonic Missions

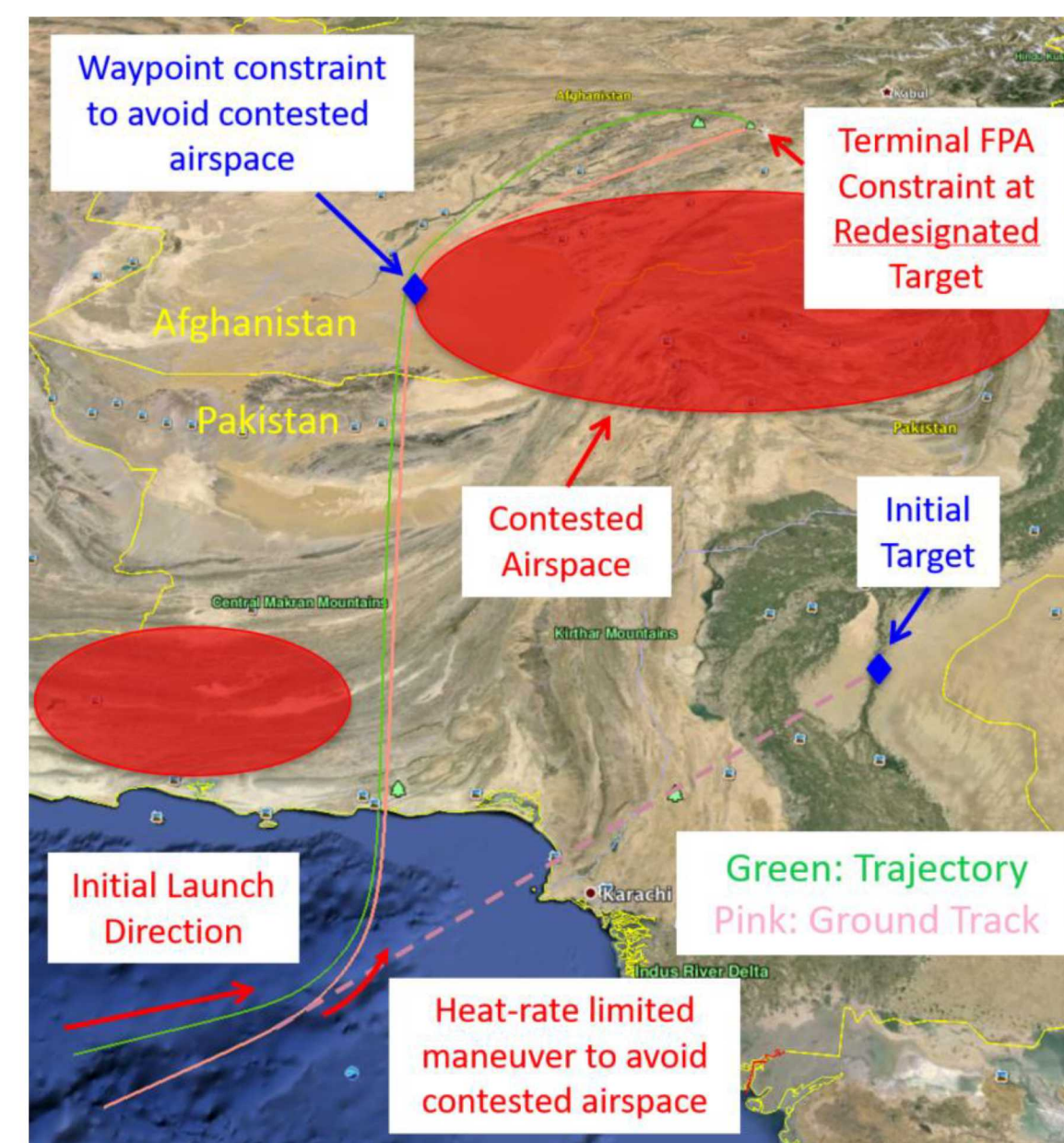
Sandia PI: Michael Grant / AA PI: Maruthi Akella, Aerospace Engineering, University of Texas at Austin



## Objective

Rather than flying preprogrammed trajectories, create a **rapid** and **reliable** “full flight” **guidance algorithm** that can update the desired flight path in an **optimal manner** based on external factors or changes to the mission

- Combine classical optimization methods for dynamical systems with machine learning



**Test** classical optimization/machine learning algorithm on representative flight hardware

## Approach

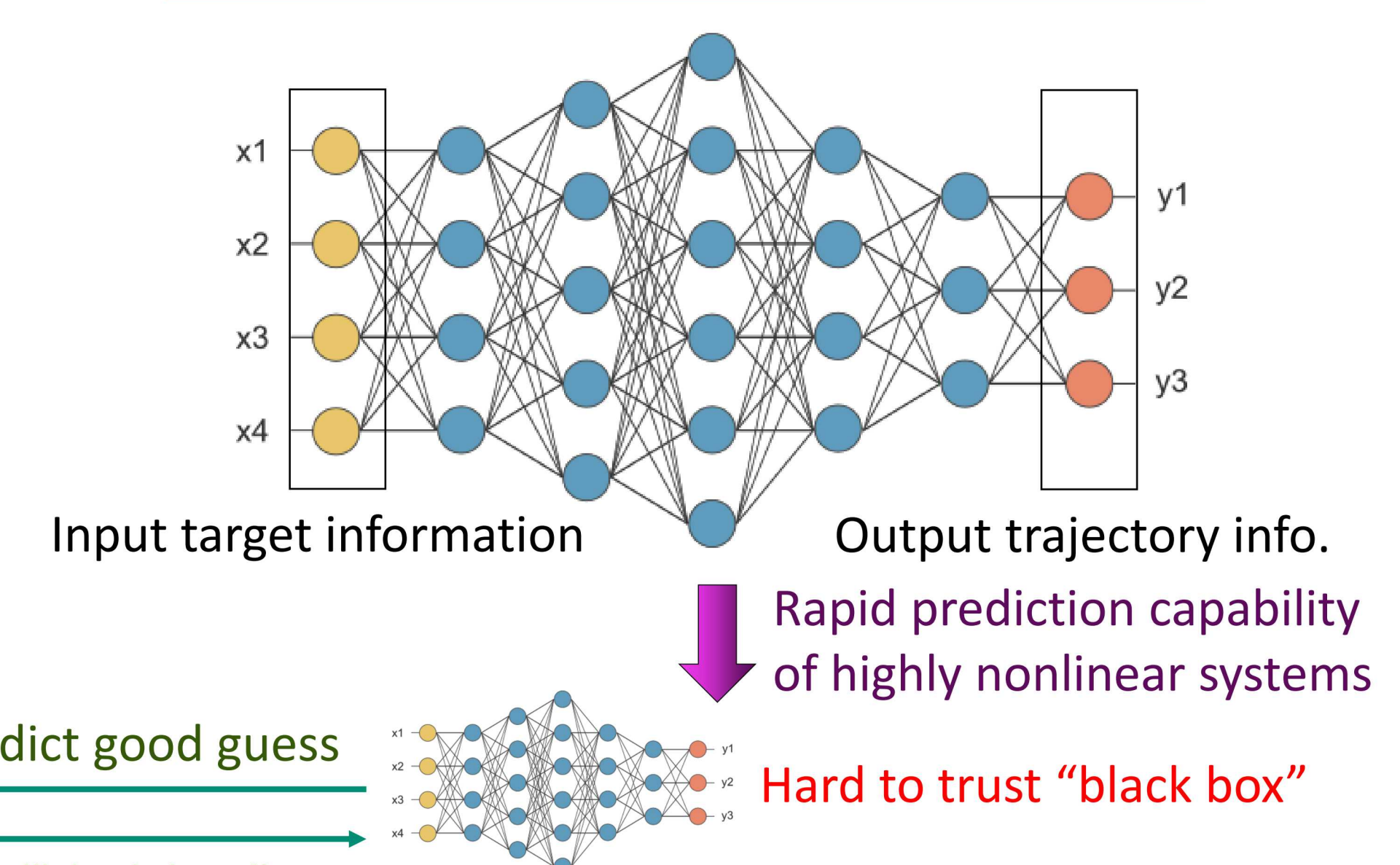
### Classical Physics-Based Optimization

IC → Target  
Lots of information along one trajectory  
Encode all trajectory information using variational calculus

Analytic trajectory data compression  
Hard to construct good initial guess to converge

Use DNN to predict good guess  
Use analytics to trust “black box”

### Modern Big Data/Machine Learning



- Expected to significantly reduce or eliminate iteration** associated with traditional guidance techniques
- Partner with Maruthi Akella (Aero. Eng. at UT Austin)** on numerical stabilization techniques

## Challenges

**Converging** to solutions, even with intelligent initial guesses from machine learning algorithms, **can be challenging**

- High sensitivities
- Numerical difficulties

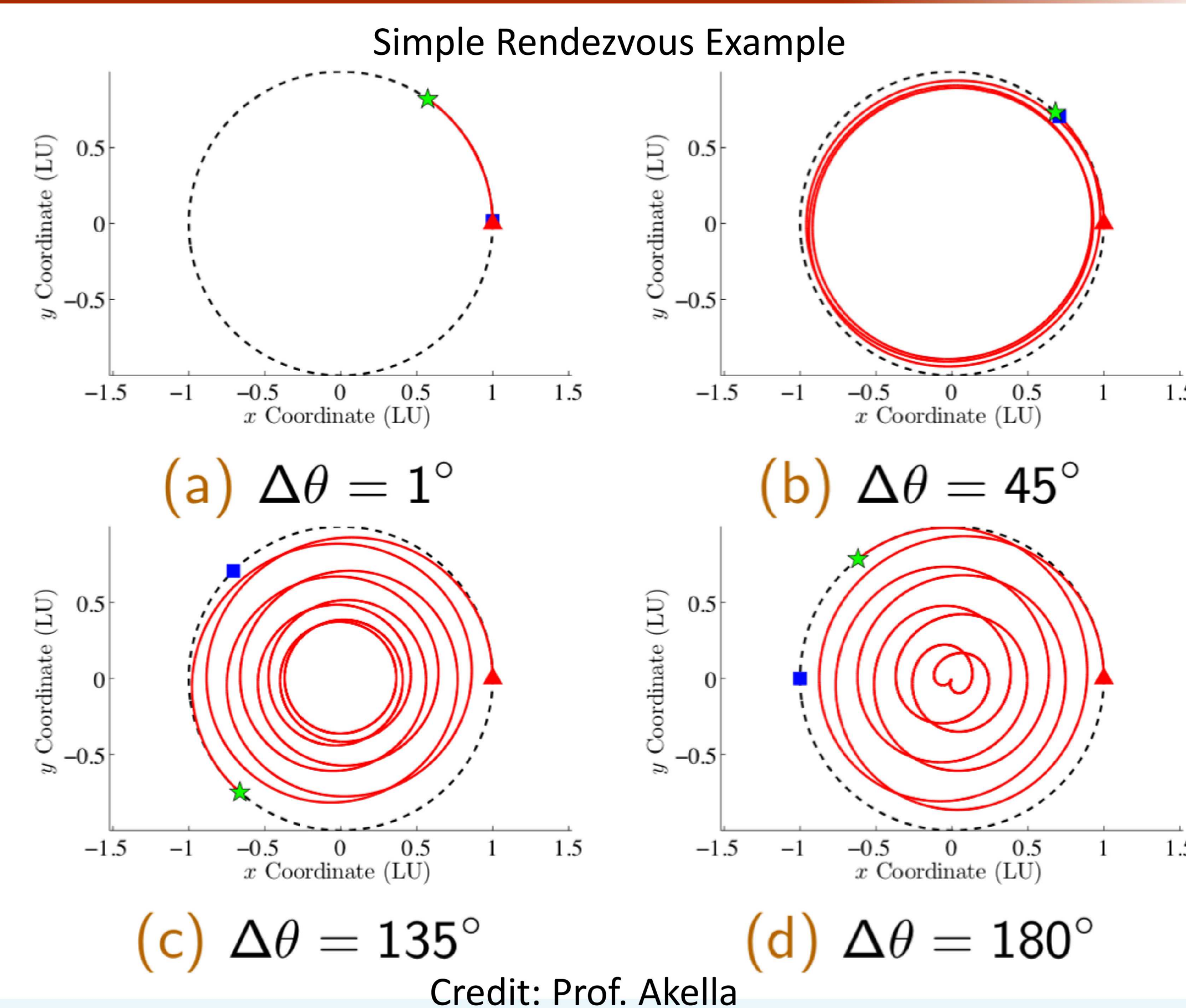
Research by Prof. Maruthi Akella has focused on **numerical stabilization techniques** that will **improve the robustness** of the overall process

- Research expected to significantly **increase the confidence** of **converging to solutions** during flight
- Approach may also **increase speed of convergence** to solutions

## Results

Prof. Akella’s research has shown to be **effective for orbital rendezvous applications**

- High sensitivities** exist due to **long flight times** associated with low thrust
- Hypersonic applications** also exhibit **large sensitivities** due to the large forces acting on the vehicle
  - Hypothesis** that **similar benefits** shown in rendezvous applications **can be leveraged** for hypersonic applications



## Partnerships

Prof. Akella’s research includes complimentary work in **finite time Lyapunov controllers** that **guarantee convergence** to feasible solutions

- Can serve as a **backup** to construct solutions in the **unlikely event** that the **numerical stabilization technique fails**
- Backup solutions expected to be **similar to optimal solutions** obtained from machine learning
  - Serve as **temporary feasible trajectories** until primary method reestablishes trajectory solutions



# Sandia National Laboratories

## Effects of *Vampirovibro chlorellavorus* on *Chlorella sorokiniana* DOE 1044 & 1116

Lauren C. Atencio, Tyler J. Hipple, Danae Maes, Jerilyn A. Timlin



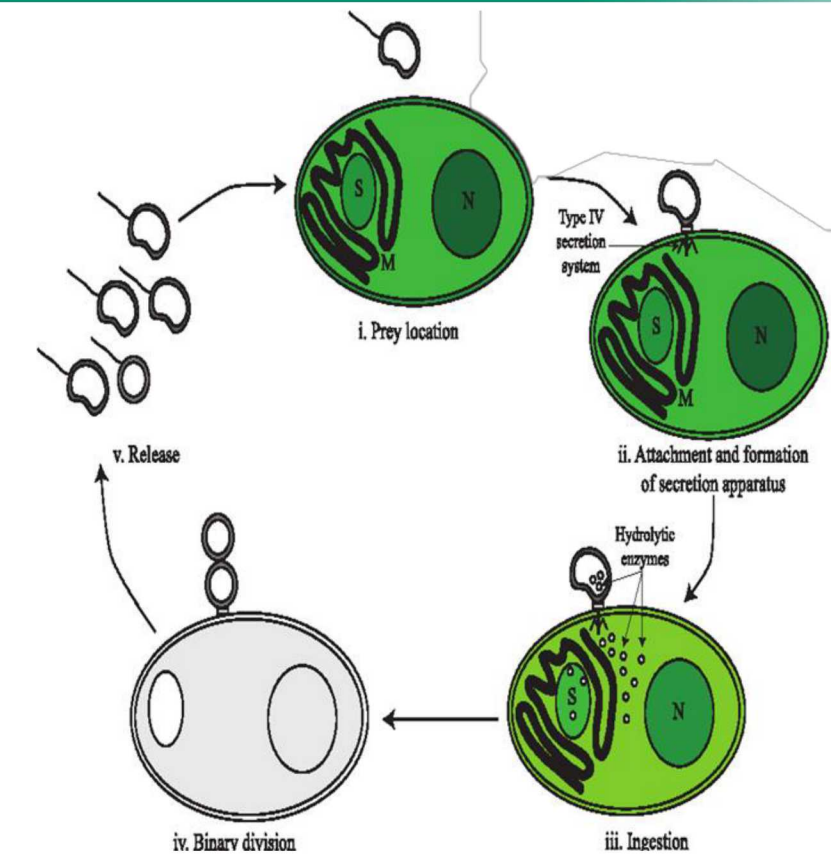
### Introduction / Motivation

#### *Vampirovibro chlorellavorus* (Vampiro):

- Gram-negative predatory bacterium
- Vampiro* infects *Chlorella* family of algae.
- Infection causes discoloration and clumping.

#### PAM (Pulse Amplitude Modulation)

- Uses pulses of light to obtain fluorescence measurements.
- Useful for assessing health of algae cultures by quantifying their photosynthetic states.



#### How *V. chlorellavorus* Attacks

- Rod-shaped bacterium attaches to the algal cell wall.
- Cell wall ruptures & bacteria enters the periplasmic space.
- Inner contents are dissolved in a matter of days.

### Approach

Infection of saltwater strains of *Chlorella* by *Vampiro* had not been validated in a laboratory setting. We challenged two saltwater strains of *C.sorokiniana* (DOE1044 and DOE1116) with *Vampiro*. The saltwater strains became infected with *Vampiro* over a two week period.

#### Experimental setup

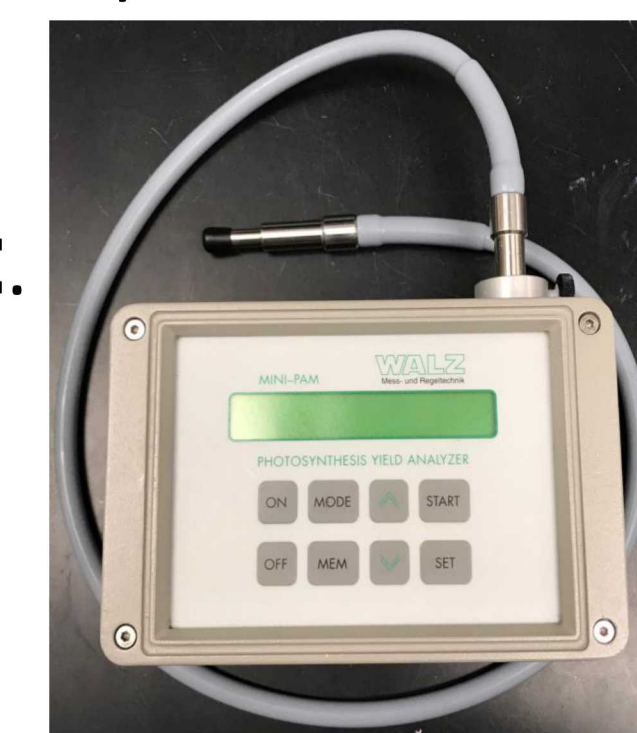
- 50mL uninfected cultures were inoculated using C1V1=C2V2 calculations of algae and media at an optical density of 0.1.
- Replicates one and two of infected cultures contained 40mL of *C. sorokiniana* with 10mL of *Vampiro* extract.
- Infected replicate three contained 30mL of *C. sorokiniana* with 20 mL of mature *Vampiro* cultures.

#### Culture Conditions

- Kept on benchtop shaker units at 140 RPM under 100  $\mu$ M of light at around 23  $^{\circ}$ C.
- Infected and uninfected were kept in separate labs.

#### Data Collection

- Optical density, PAM, light intensity, and camera photos taken every 24 hours.
- Transmitted light microscopy taken every 3 days.



### Challenges

- All uninfected cultures were growing at slower rates than the infected cultures.
- This slow growth rate could be due to shocking the cultures, but this has not been confirmed.
- The cultures were not kept in incubators resulting in variable light and temperature conditions.

### Results

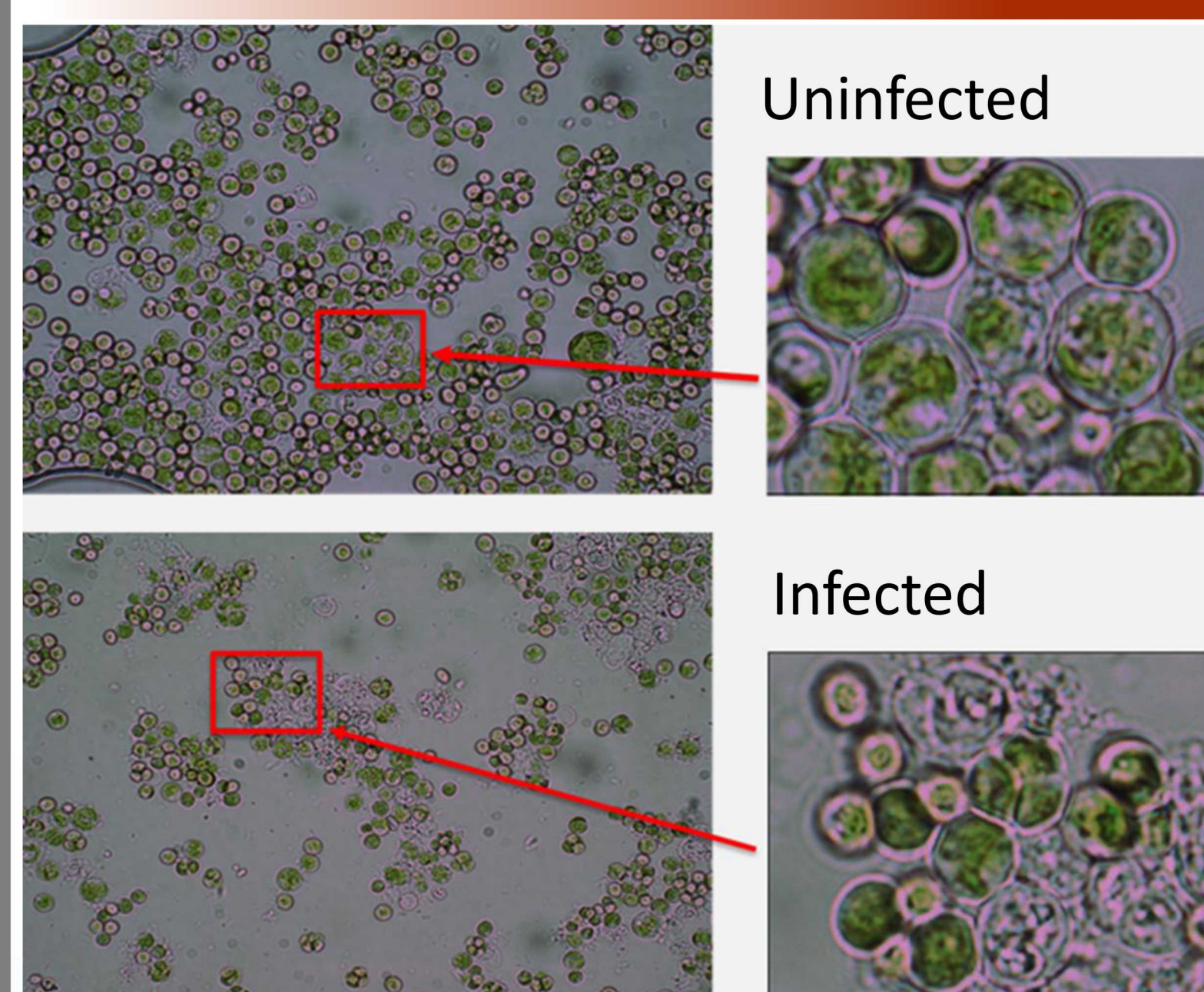
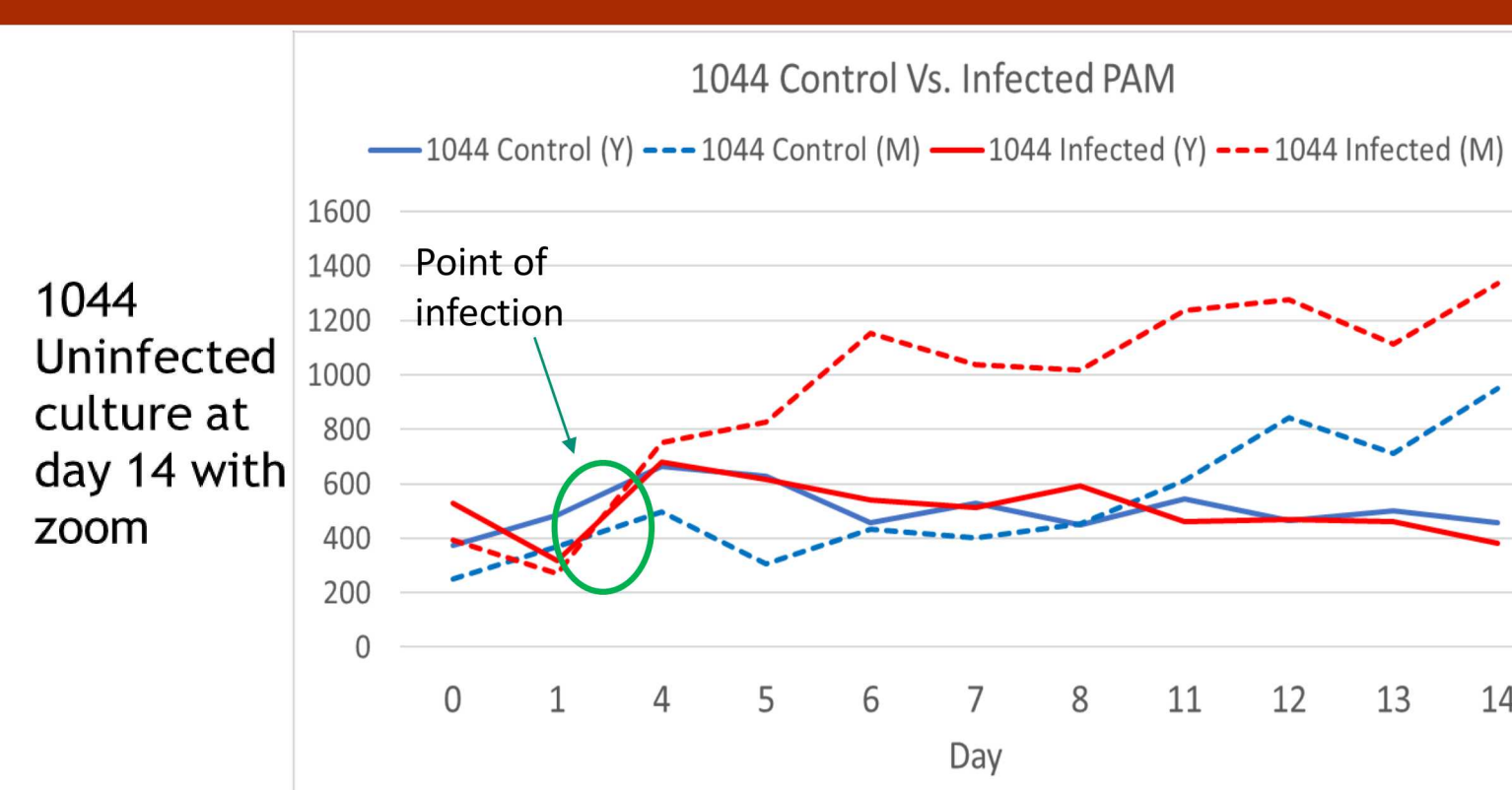
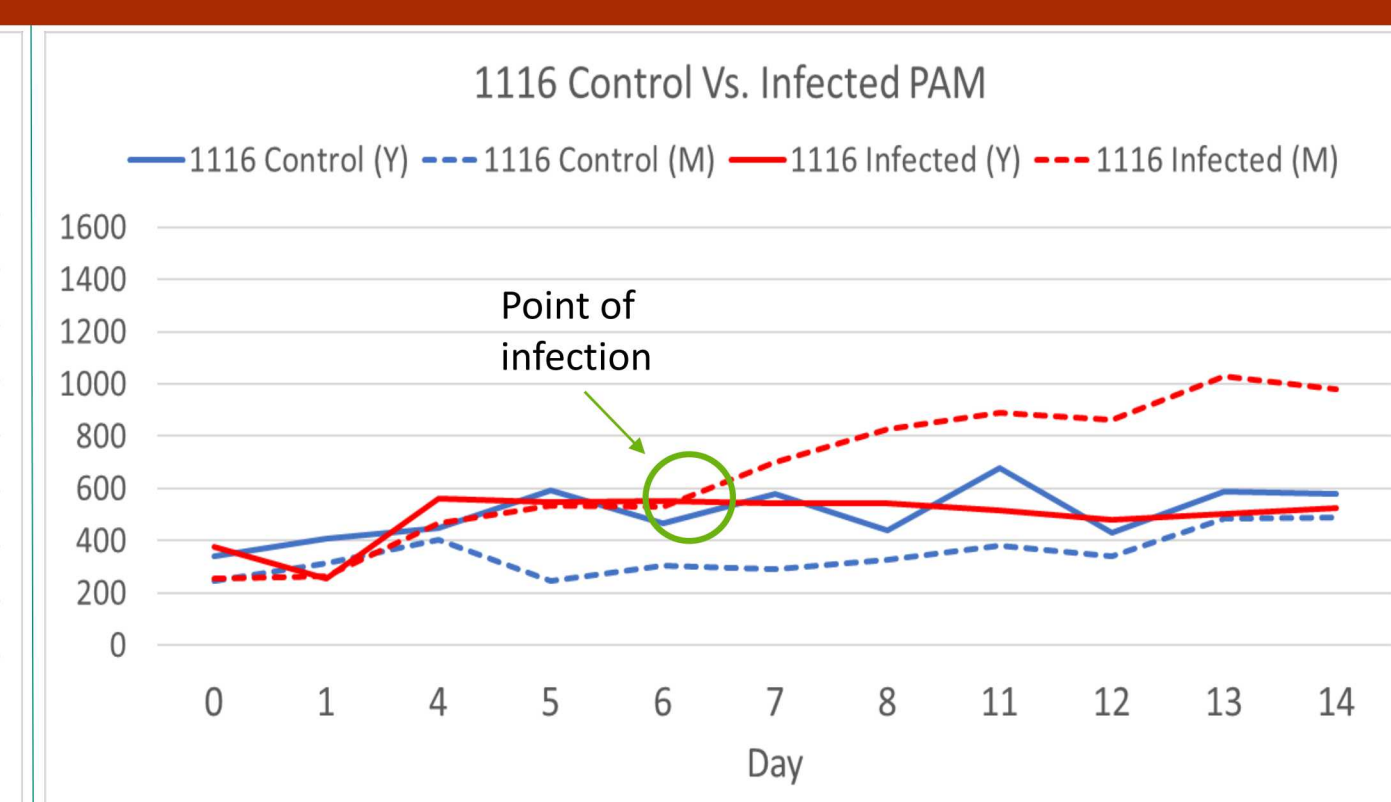
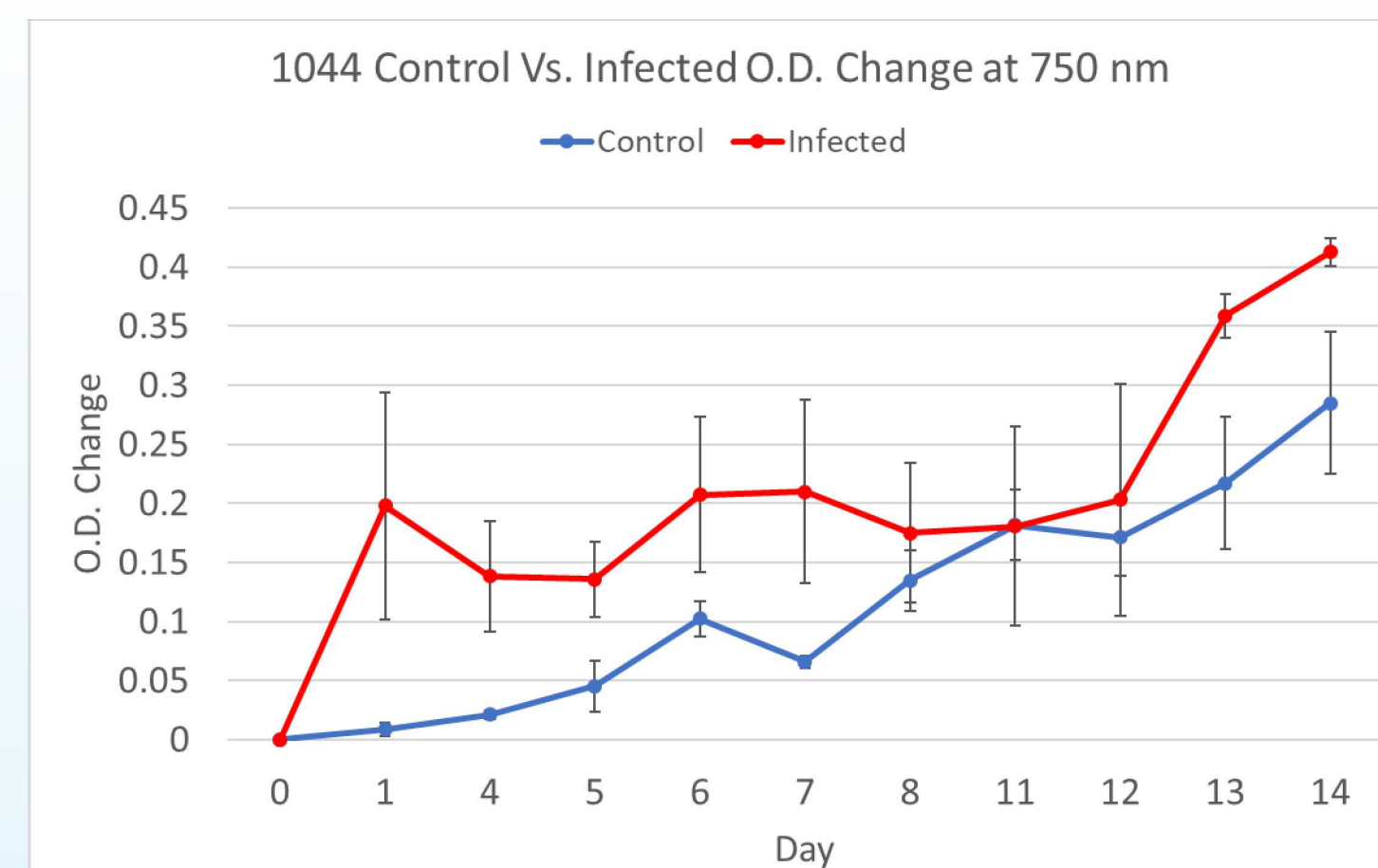


Figure 1: Comparison of infected and uninfected *C.sorokiniana* DOE 1044 (Top left) Transmitted light microscopy of uninfected Vs. Infected on day 14. (Top right) PAM measurements over a 14 day period; infection is indicated when the M and Y lines cross. (Bottom right) Optical density readings over a 14 day period.



1044 Uninfected culture at day 14 with zoom



1116 Uninfected culture at day 14 with zoom

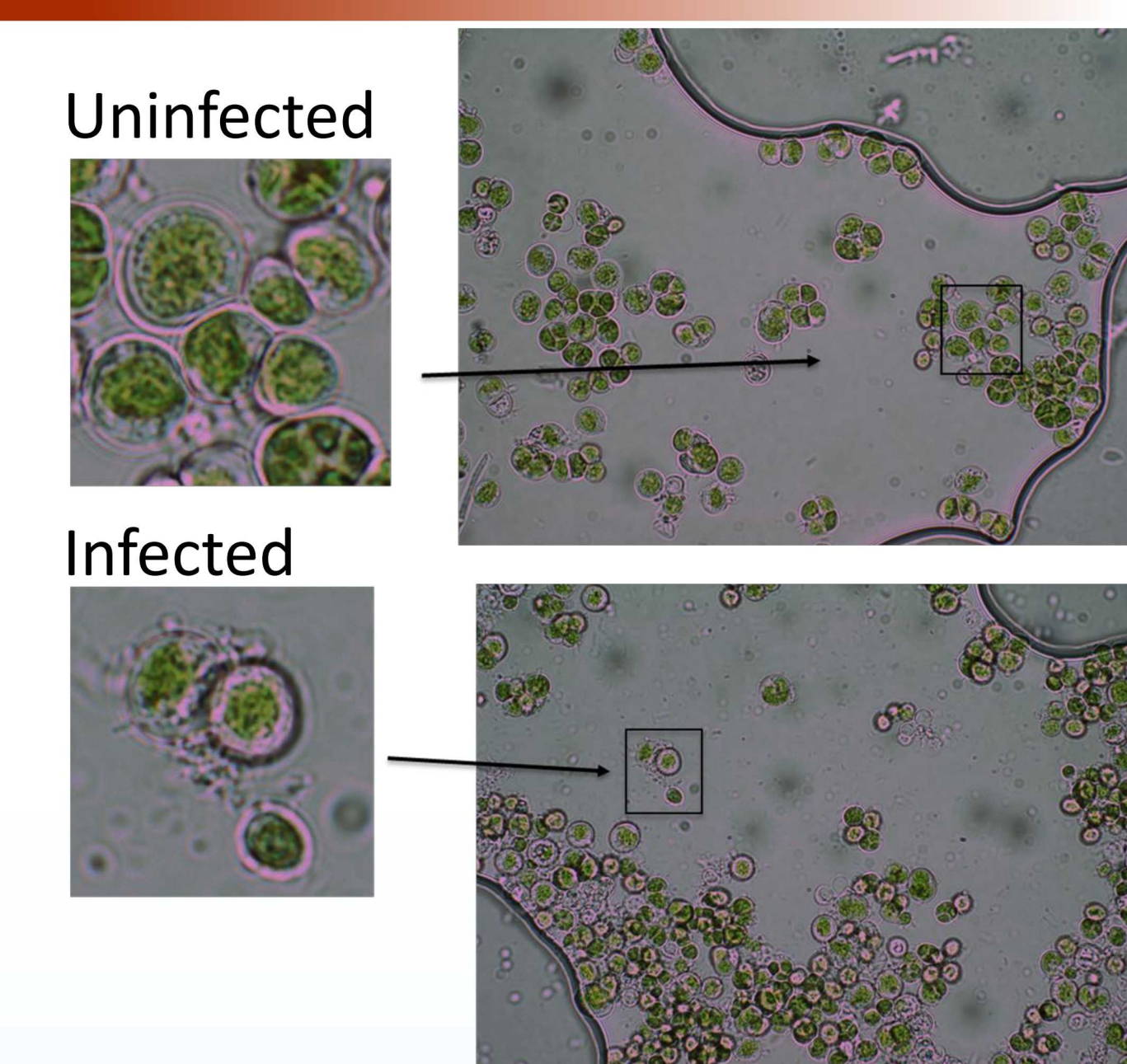
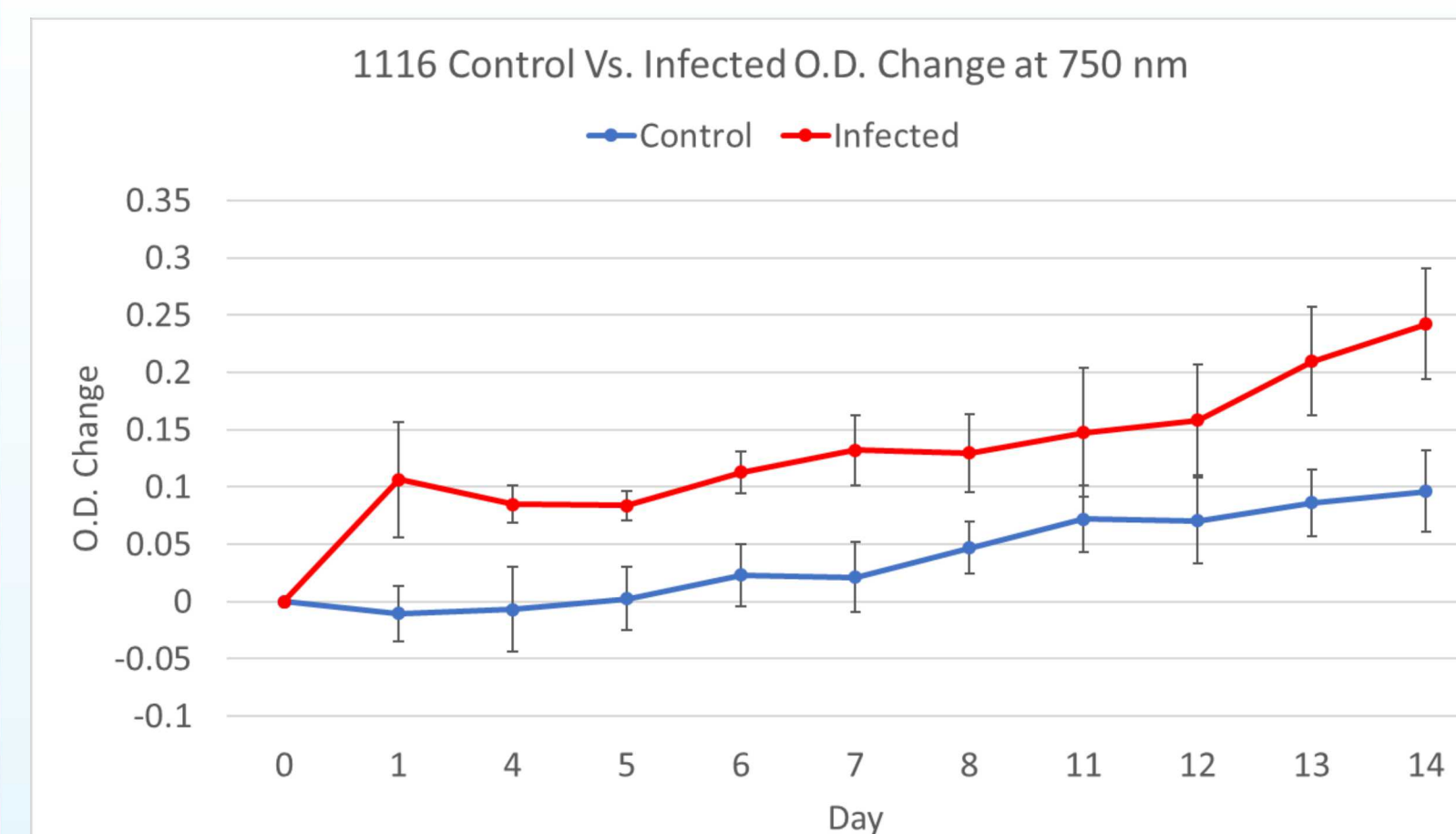


Figure 2: Comparison of infected and uninfected *C.sorokiniana* DOE 1116 (Top right) Transmitted light microscopy of uninfected Vs. Infected on day 14. (Top left) PAM measurements over a 14 day period; infection is indicated when the M and Y lines cross. (Bottom left) Optical density readings over a 14 day period.

### Future Work / Acknowledgements

Repeating the experiment with acclimated cultures in incubators to reduce assumed initial shock phase and keep conditions constant

This work was completely funded by BioEnergy Technologies Office (BETO) through the DISCOVER project.



#### References:

- V. Chlorellavorus* infection diagram from Soo, R.M. et. al., 2015, Peer J



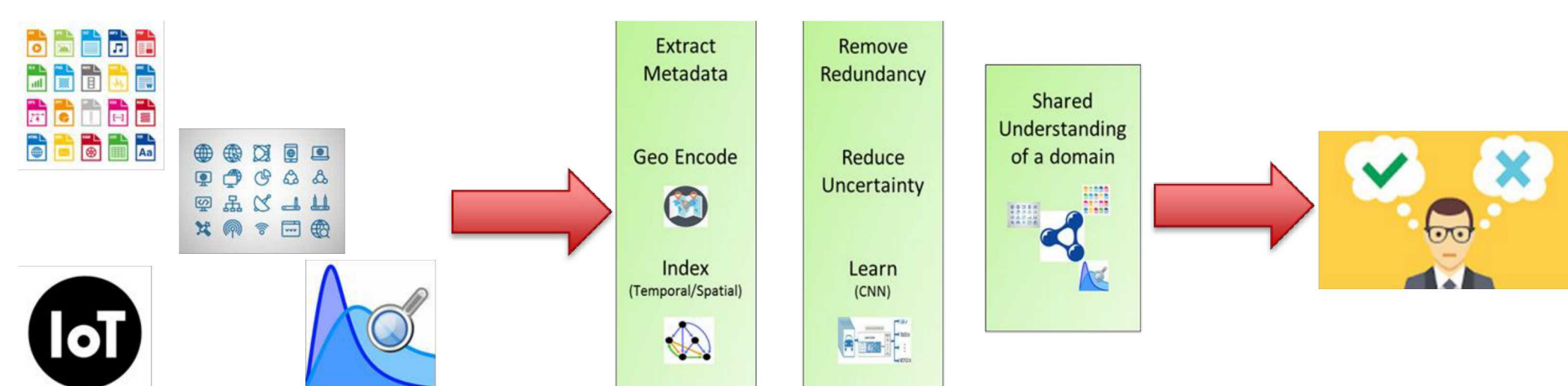
# Sandia National Laboratories Turning Big Data into Actionable Intelligence

Tian Ma, Rudy Garcia, Forest Danford, Laura Patrizi, Jenny Galasso, Tim Draelos, Thushara Gunda, Jason Loyd, Otto Venezula, Wesley Brooks

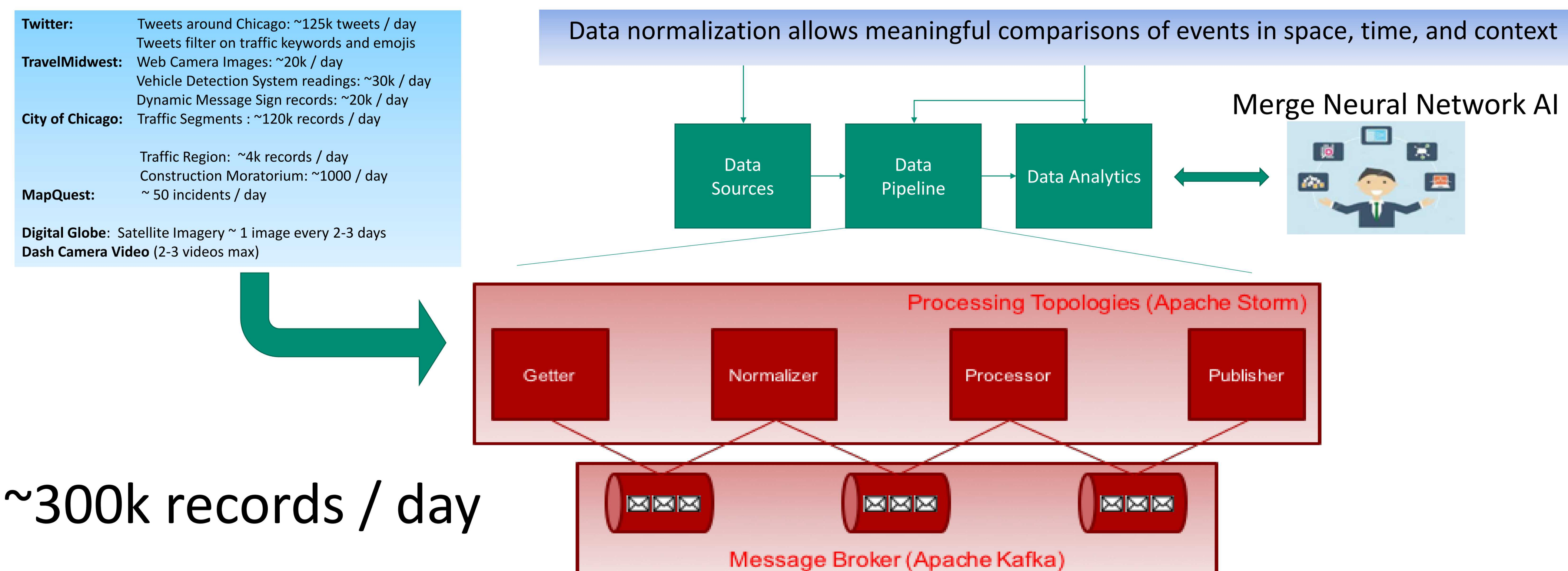


## Introduction / Motivation

- Significant increase in amount of digital information
  - “90% of the data in the world today has been created in the last two years alone, at 2.5 quintillion bytes a day” – IBM Marketing Cloud (December 2016)
  - There are more data than human can analyze
- Data are “Gold” only if you can uncover an insight
- Quickly turn real-time streaming data from variety of sources into actionable insights that enables decision makers from political leaders to field commander to take appropriate action when faced with security threat



## Approach: Big Data Actionable Intelligence Architecture



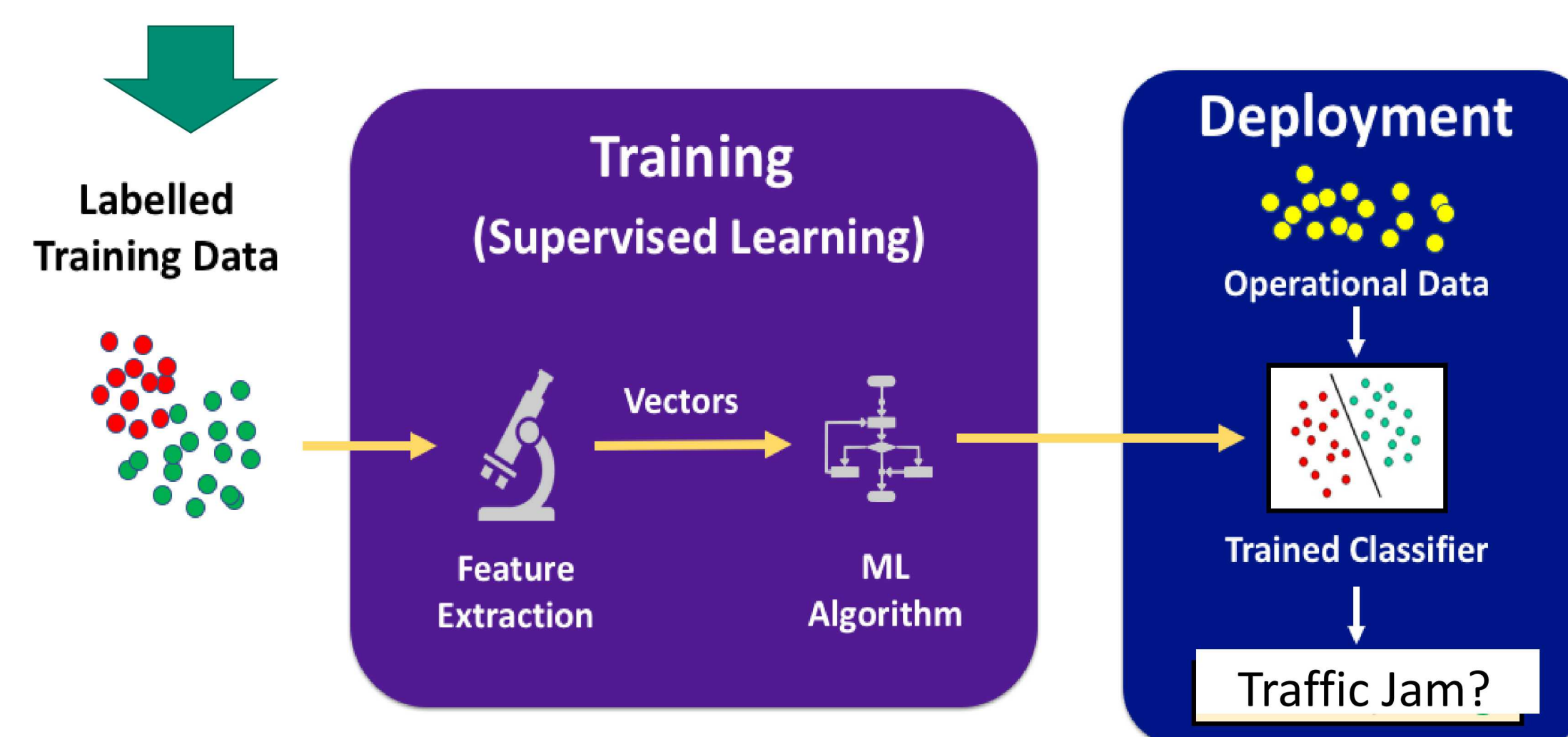
## Challenges

- Scalability
  - Storage
  - Computing
  - Number of Data Sources
- Uncertainty quantification of information in decision making

## Results

### Exemplar Problem: Near-Real-Time Traffic Estimation Using Variety of Sources (without using GPS crowd sourcing (i.e. Google Maps))

Camera images and Tweets



Merge Neural Network AI

Confusion matrix

|      | cat1 | cat2 | cat3 |
|------|------|------|------|
| cat1 | 0.87 | 0.07 | 0.05 |
| cat2 | 0.17 | 0.83 | 0.00 |
| cat3 | 0.00 | 0.33 | 0.67 |

Cat1: No Traffic Impact  
Cat2: Low Traffic Impact  
Cat3: High Traffic Impact

## Partnerships

University of Illinois



Illinois Applied Research Institute



Sandia National Laboratories





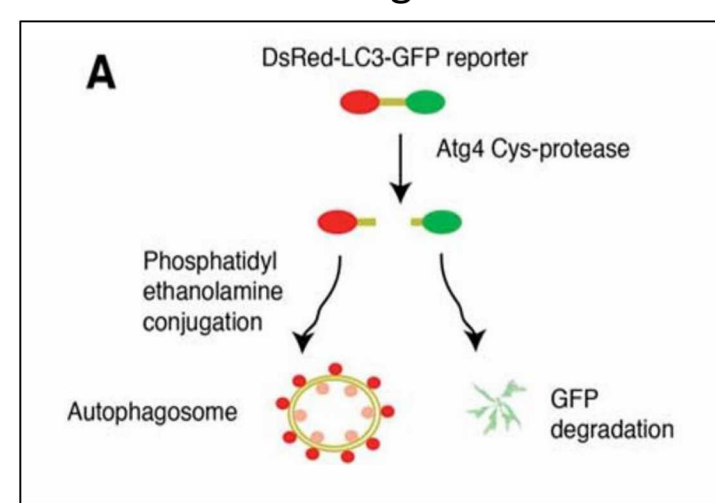


### Introduction / Motivation

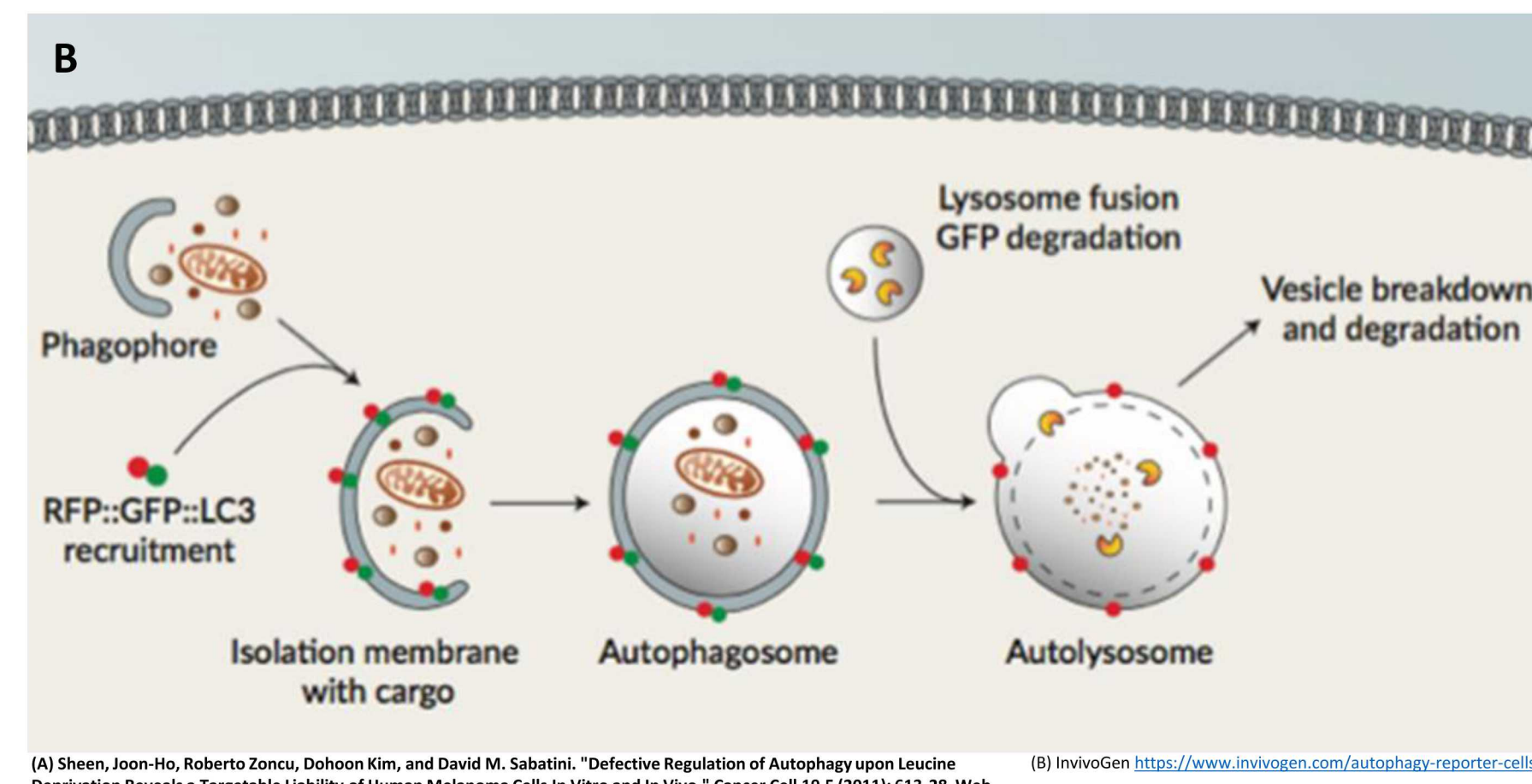
- Autophagy is a regulated process active in all human cells. It eliminates damaged organelles, removes toxic aggregated proteins, and acts as an autonomous defense against intracellular pathogens.
- Previous studies<sup>[1]</sup> have displayed autophagy as having potential for therapeutic host-targeted control of mycobacterial infections through autolysosomal killing. This would limit the generation of antimicrobial peptides and potentially dangerous inflammation.
- Our goal is to demonstrate autophagy stimulation as a useful adjunct to antibiotic therapy in mycobacterial infections with the potential to limit drug resistant strains.

<sup>[1]</sup> Bradfute, S. B., et al. (2013). "Autophagy as an immune effector against tuberculosis." *Current Opinion in Microbiology* 16(3): 355-365.

We have used an engineered HeLa DsRed-LC3-GFP autophagy reporter cell line to quantify drug-induced autophagy induction → fluorescence change



Increase in RFP with a concurrent decrease in GFP



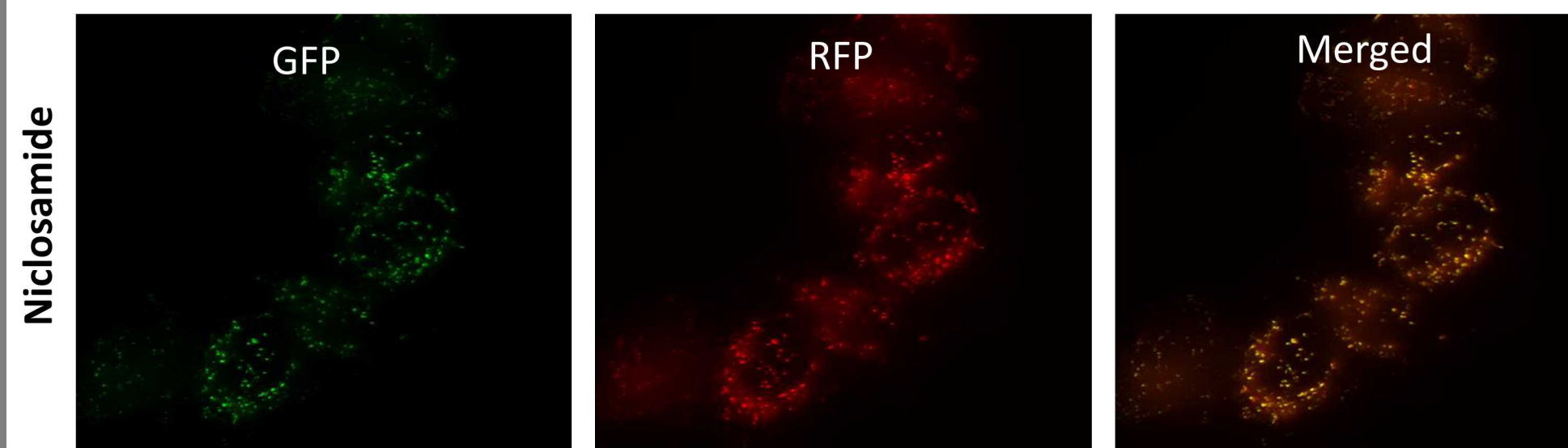
(A) Sheen, Joon-Ho, Roberto Zou, Dohoon Kim, and David M. Sabatini. "Defective Regulation of Autophagy upon Leucine Deprivation Reveals a Targetable Liability of Human Melanoma Cells in Vitro and in Vivo." *Cancer Cell* 19.5 (2011): 613-28. Web. (B) InvivoGen <https://www.invivogen.com/autophagy-reporter-cells>

### Approach

Confocal fluorescence imaging and a HeLa cell line engineered to express an autophagy marker was utilized to quantify the effectiveness of 4 approved drugs that stimulate autophagy. We then tested the efficacy of a multimodal-treatment strategy consisting of a drug and a TB antibiotic in RAW 264.7 cells infected with *M. bovis*, a TB surrogate.

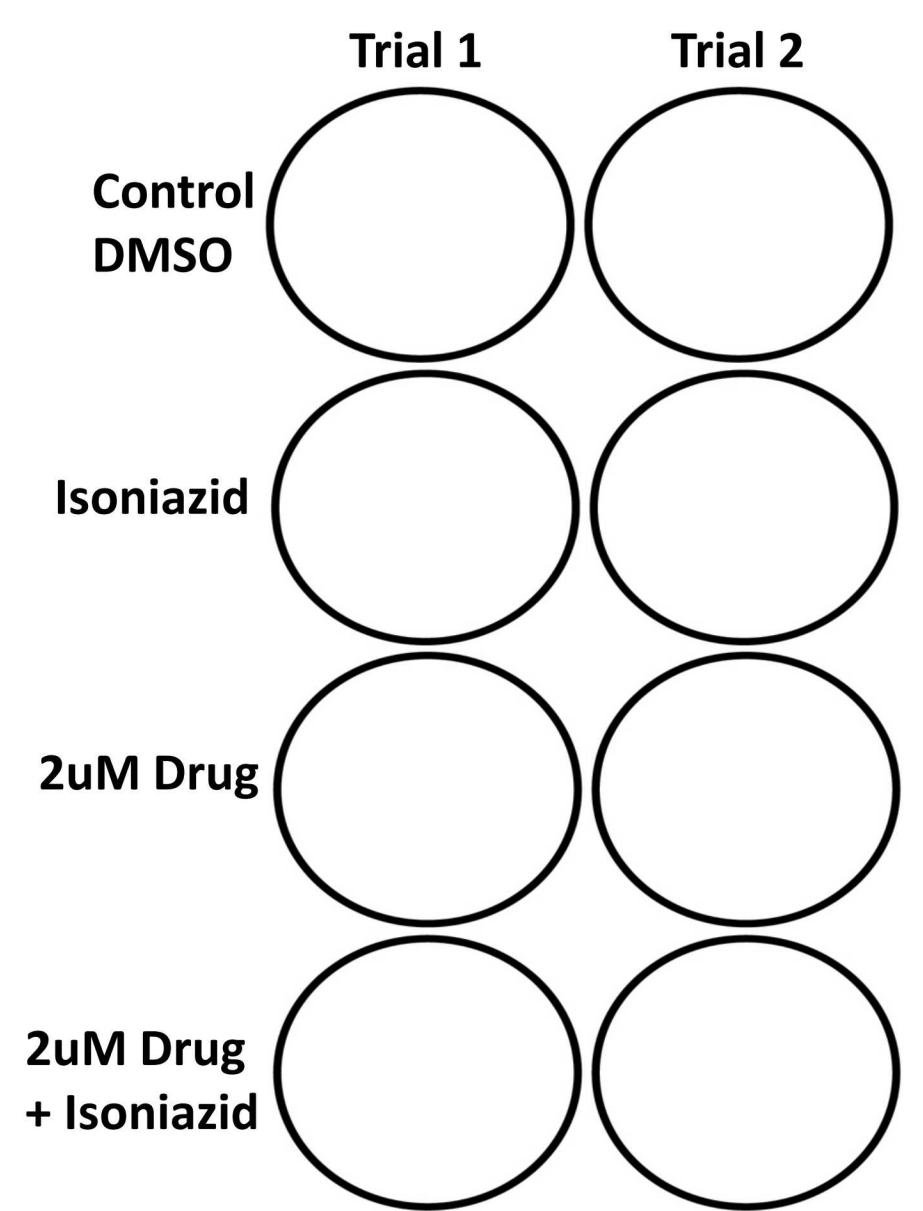
#### Measuring Autophagy Induction in Single Cells

- HeLa RGL1 cells were cultured in DMEM-10 with 500 ug/ml G418 antibiotic at 37°C and 5% CO<sub>2</sub>.
- 2uM autophagy stimulating drug (niclosamide, rapamycin, flubendazol, or bromhexine HCl) was added, incubated 4 hrs.
- Images were collected at 4-6 hours post addition of drug using an inverted fluorescence microscope (Olympus IX-71).



#### Testing Efficacy using a TB Surrogate

- RAW 264.7 (mouse macrophage) cells were infected with *M. bovis* at ~MOI10 for 1 hour at 37° C and 5% CO<sub>2</sub>.
- 2uM of autophagy stimulating drug and/or antibiotics was added, incubated 4 hrs.
- Cells were lysed using 0.1% Triton X-100 and mycobacterium collected from the lysate using centrifugation.
- M. bovis* was washed, diluted, and plated onto Middlebrook 7H9 agar plates.
- Colonies were allowed to grow at 37°C for 10-14 days.
- CFU count was used to determine the efficacy of the drug/antibiotic combination.



### Challenges

#### Single-Cell Analysis of Autophagy Stimulating Drugs

##### Step 1: Merge and Flatten Tiffs

In-house written software to merge and flatten the 14-stack tiffs to easily identify cells in the image.

##### Step 2: Segment Image to Identify Individual Cells

Utilize in-house written software, CellFinder, to identify the outline of individual cells in the images.

##### Step 3: Identify and Quantify Puncta in Both GFP and RFP Channels

In-house written software, BatchBiophagyCell, quantifies the number, intensity, and area of the puncta in each channel.

- Goal:** Identify number, area, and intensity of RFP and GFP puncta in individual cells
- Previous studies<sup>[1]</sup> have measured autophagy induction by taking the average intensity over an entire image. The background intensity from multiple cells is a hindering variable producing less accuracy. 14 cell-based metrics were calculated and 3 selected to compare autophagy:
  - Average GFP intensity
  - RFP:GFP average puncta intensity ratio
  - RFP:GFP average puncta area ratio

<sup>[1]</sup> Chauhan S, et al. (2015) Pharmaceutical screen identifies novel target processes for activation of autophagy with a broad translational potential. *Nature Communications* 6, doi:10.1038/ncomms9620

### Results

#### Autophagy Induction in Single Cells

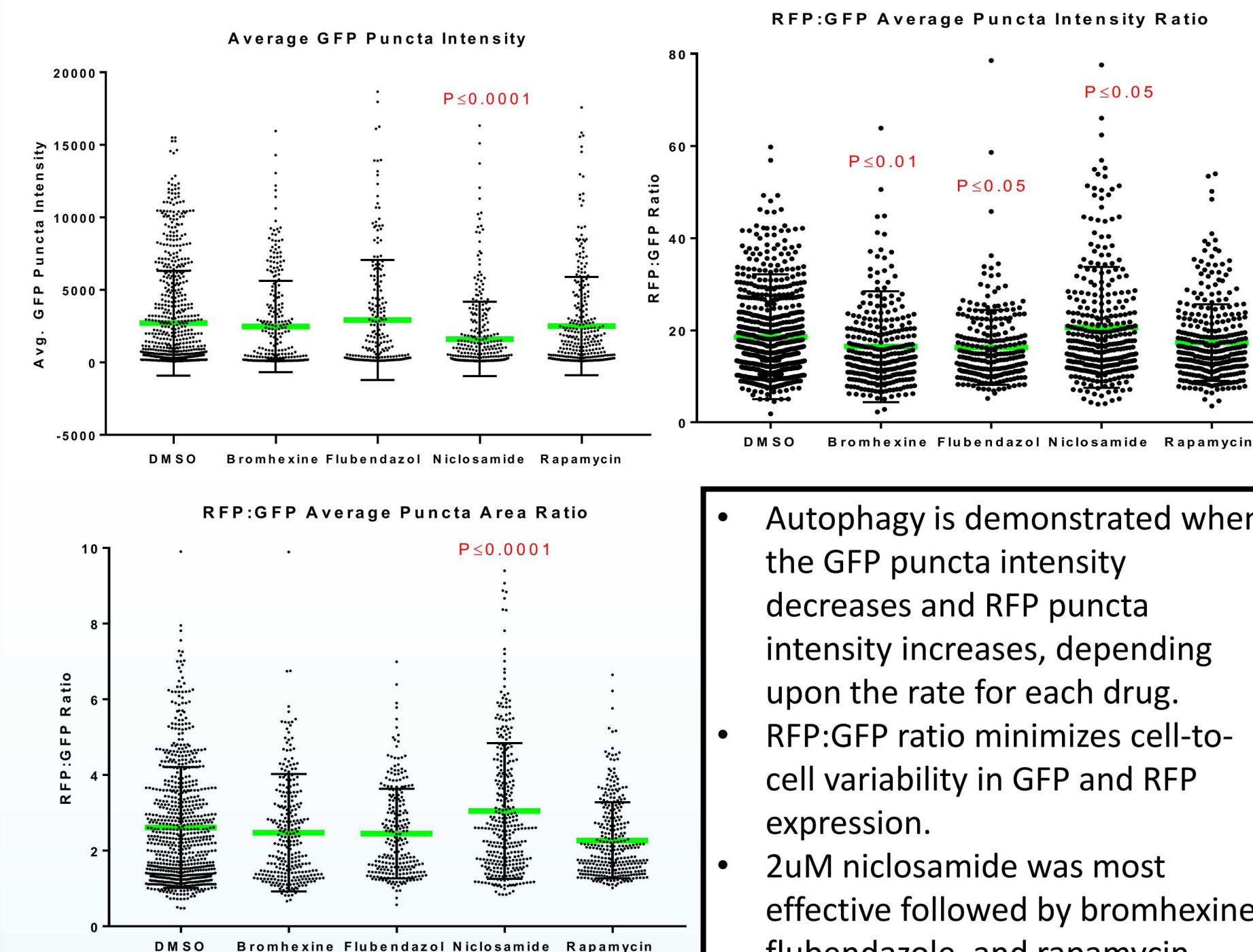


Figure 1: Top three metrics of approved drugs to determine efficiency of autophagy induction based upon single cell confocal fluorescence microscopy.

Single cell analysis was completed on a 14-slice confocal stack, where each drug condition has >250 cells. Statistical significance was determined by Mann-Whitney test followed by a Dunn's multiple comparison test to compare conditions to DMSO. Errors bars represent the SD from 3 independent experiments.

#### Efficacy in a TB Surrogate

| <i>M. bovis</i> infected RAW 264.7 Clearance |                                     |                                     |                    |
|--|-------------------------------------|-------------------------------------|--------------------|
| Drug Treatment                               | Replication 1 (% Survival) MOI 3.75 | Replication 2 (% Survival) MOI 0.75 | Average % Survival |
| DMSO   | 100                                 | 92                                  | 96                 |
| Isoniazid                                    | 34.7                                | 94.7                                | 64.7               |
| Niclosamide                                  | 240                                 | 70.7                                | 155.35             |
| Flubendazol                                  | 60                                  | 50.3                                | 55.15              |
| Bromhexine HCl                               | 293                                 | 73.3                                | 183.15             |
| Rapamycin                                    | 20.7                                | 46.7                                | 33.7               |
| Isoniazid + Niclosamide                      | 6                                   | 49.3                                | 27.65              |
| Isoniazid + Bromhexine HCl                   | 19.3                                | 62.7                                | 41                 |
| Isoniazid + Rapamycin                        | Contaminated                        | 45.3                                | 45.3               |
| Isoniazid + Flubendazol                      | 11.3                                | 57.3                                | 34.3               |

Fig. 2. RAW 264.7 infected with *M. bovis* to determine killing synergy based upon a multi-modal drug combination.

Percent survival of *M. bovis* in each replication. In replication 1, both niclosamide and bromhexine displayed >100% survival which could be attributed to *M. bovis* aggregates that form resulting in an uneven distribution in the CFU assay. In replication 1, Rapamycin was uncountable due to contamination.

- Each autophagy drug enhanced isoniazid killing.
- 2uM niclosamide in combination with 0.4ug/mL isoniazid was the most effective followed by flubendazol, bromhexine, and rapamycin combinations.

### Future Work / Partnerships

- Build on this model to characterize a set of new, proprietary antibiotics and autophagy stimulants. We will look at:
  - Autophagy stimulation efficacy
  - Antibiotic potency vs. isoniazid
  - Potential synergy of autophagy stimulants with isoniazid or the new antibiotics
- Identify potential drug binding sites and interactions within a human cell line.



Funding was provided by the NMSBA Program





## Introduction / Motivation

TracerFIRE is a Forensic and Incident Response Exercise program simulating a set of corporate-level cyber-attacks which produce realistic forensic artifacts.

For TracerFIRE to be effective, a realistic network infrastructure that can be infected with malware and allow for the collection of forensic artifacts is necessary. Due to hardware constraints and a portable platform prerequisite, the network needs to be virtualized. Additionally, the network needs to be customized for specific cyber attack killchains.

## Approach

Construct a virtualized network, to host customized virtual machines and services for the TracerFIRE scenario. These are the technologies used to accomplish this:

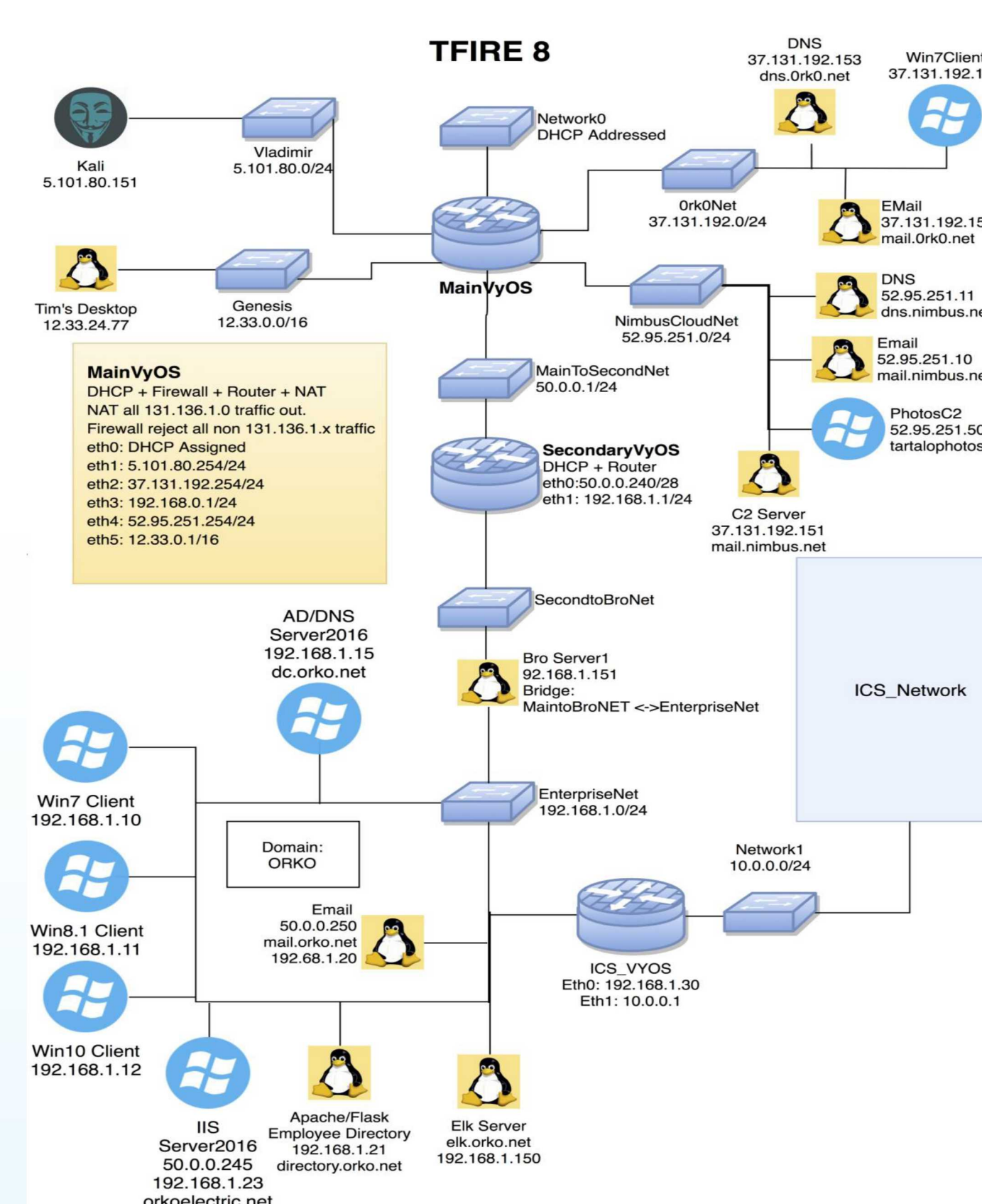
- VyOS – controls routing and manages firewalls within the network
- Xen Hypervisor – allows us to create and reproduce the virtual network across multiple hardware devices
- Bind9, Nginx, iRedMail – provides DNS, web and email services on corporate and attacker networks
- Apache/Flask - hosts the employee directory and syncs with AD using LDAP
- Ubuntu/Debian OS – hosts opensource web services
- Windows Server 2016 – hosts AD/DC, DNS, and IIS
- Bro Service – logs network traffic for analysis
- Filebeat and ELK Stack - used to forward, view, and search network logs
- Orchestration framework - emulates a realistic ICS network

## Challenges

- Manipulation of Virtual machines and Networks on the Xen Hypervisor
- Windows network administration and enterprise management
- Using the DNS within the confines of the TracerFIRE scenario
- Successful execution and simulation of attack chains on the network.

## Results

- A virtualized network that can be redeployed on multiple Xen instances
- A network infrastructure that supports the requirement of the TracerFIRE scenarios
- The network allows malware infection within a simulated corporate network



## Future Work / Partnerships

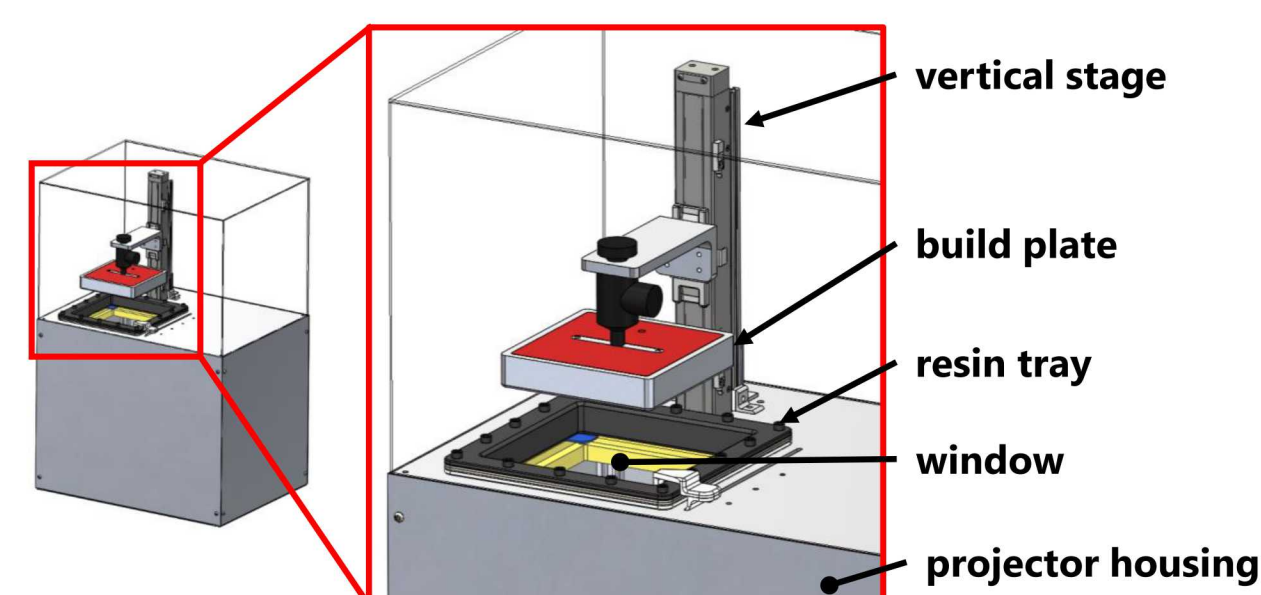






## Introduction / Motivation

Digital light processing (DLP) 3D printing can be used for manufacturing complex structures using a variety of materials, which would be nearly impossible using traditional manufacturing methods. Recent work at Sandia National Laboratories uses DLP technology for additive manufacturing of complex alumina structures, using photocurable resins loaded with micron or submicron alumina particles. These resins are printed using a DLP 3D printer to produce a "green part." The work presented here will discuss the mechanical challenges associated with printing alumina using commercially available DLP and stereolithography 3D printers, including the design of a custom DLP 3D printer to address identified mechanical challenges, thereby leading to improved print versatility and quality.

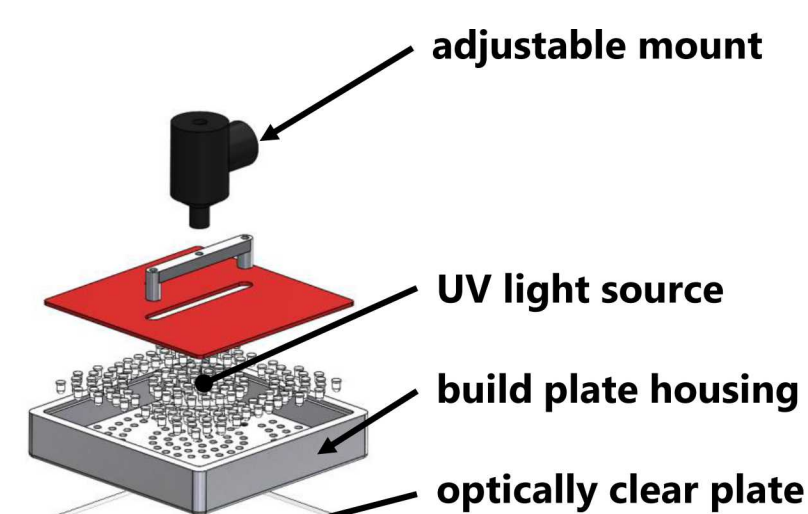


## Custom Printer

### UV Backlit Build Plate

The UV backlit build plate is intended to:

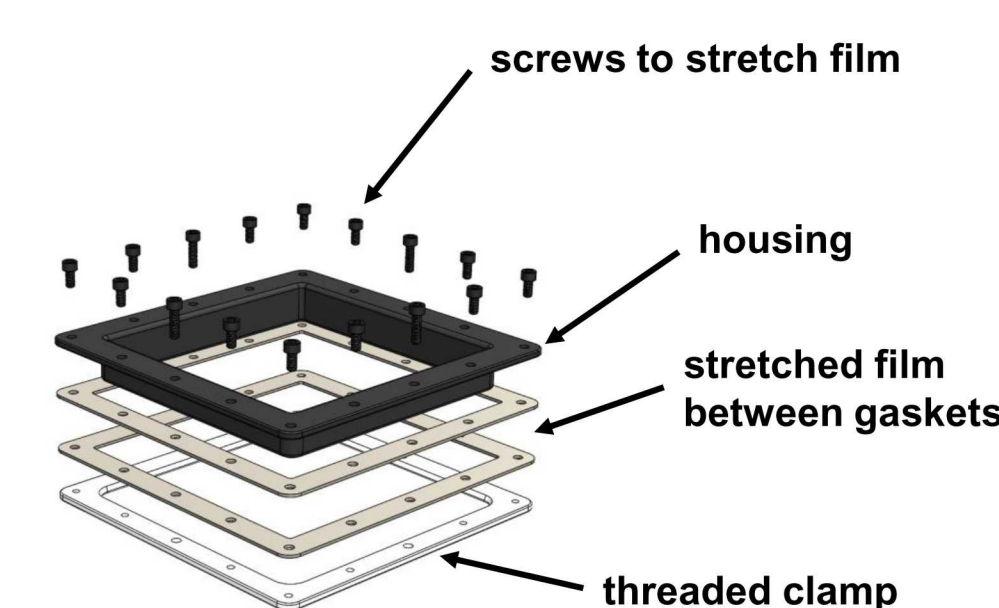
- Ensure strong build adhesion
- Provide a level surface between resin tray and build surface



### Stretched Film Resin Tray

The stretched film resin tray is intended to:

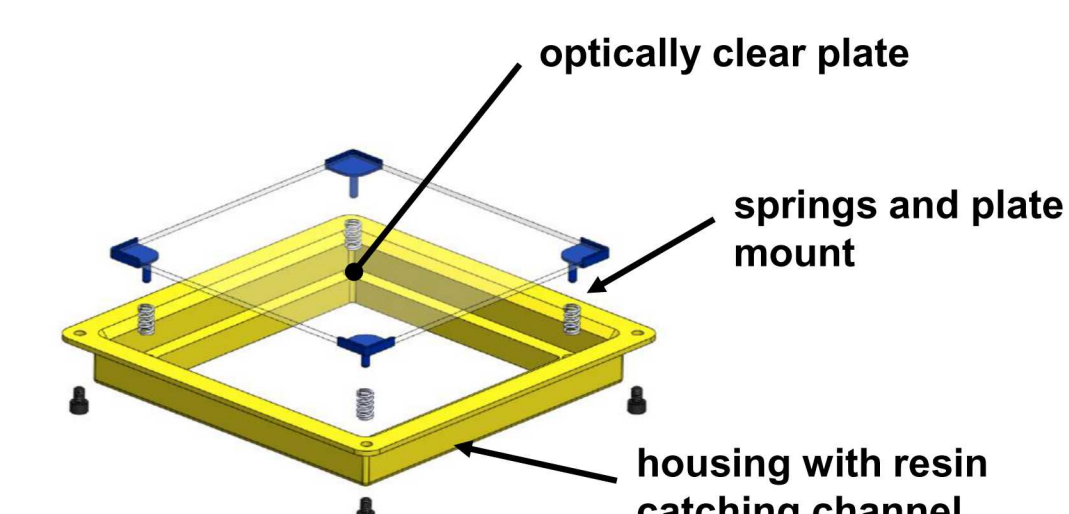
- Allow builds to slowly release from resin tray
- Reduce force on build due to separation
- Provide an optically clear, durable, surface to build from



### Glass Support Plate

The glass support plate is intended to:

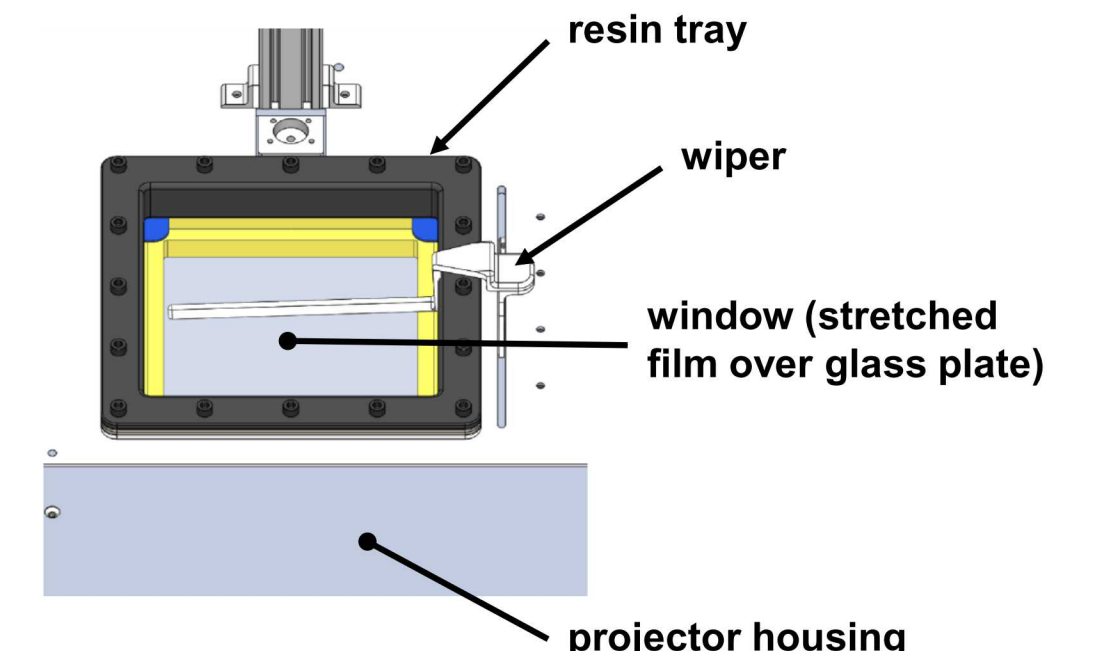
- Provide a level surface for the stretched film to come in contact with
- Provide a level surface for the build plate to allow for uniform layer thickness
- Prevent damage to print surfaces and printer



### Wiper Bar

The wiper bar is intended to:

- Remove over cured materials from active print area
- Agitate material to prevent particle settling

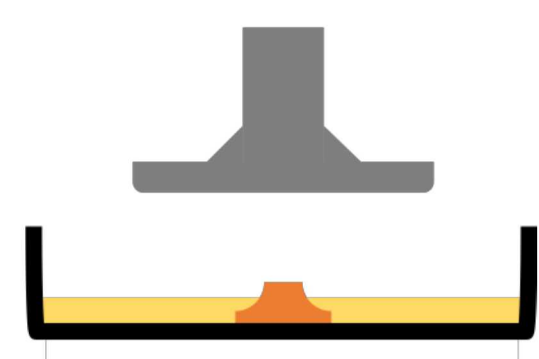


## Challenges

### Dropping

Dropping occurs when the initial layers of the build are not strongly adhered to the build plate.

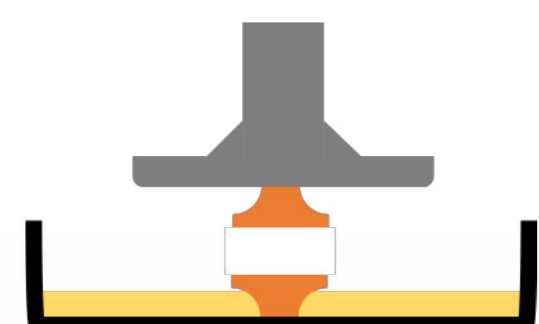
**Solution: UV-Backlit build plate**



### Splitting

Splitting occurs when the force of separation on the build is too great, resulting in a build fracture.

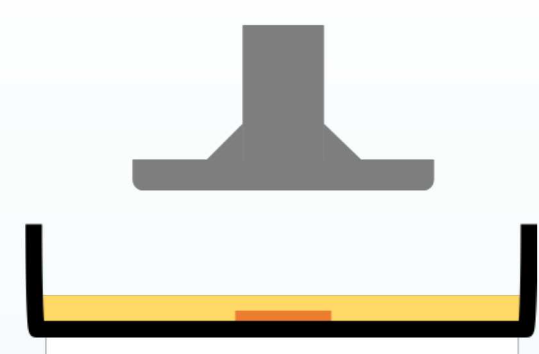
**Solution: Stretched film resin tray**



### Sticking

Sticking occurs when the build plate does not have adequate contact with the resin tray.

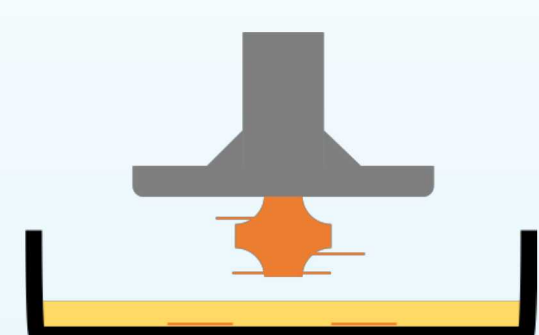
**Solution: Glass support plate**



### Flaking

Flaking occurs due to the over-curing of resin. This over-curing occurs frequently in advanced materials.

**Solution: Wiper bar**

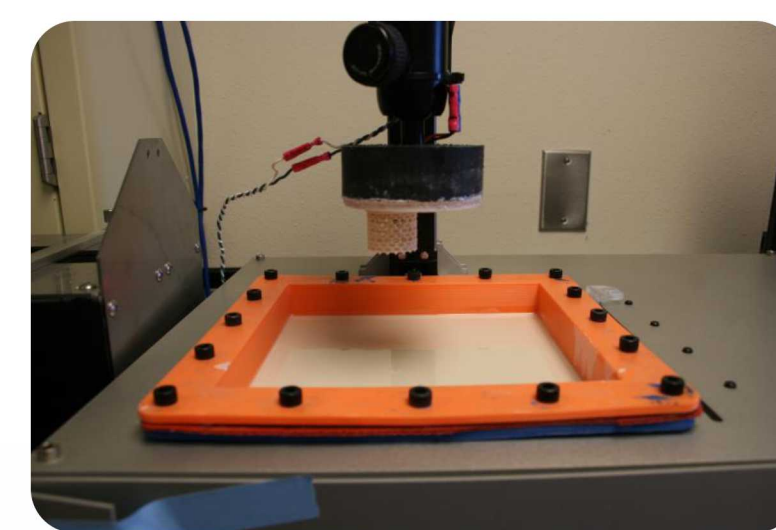


## Results

### Prototype

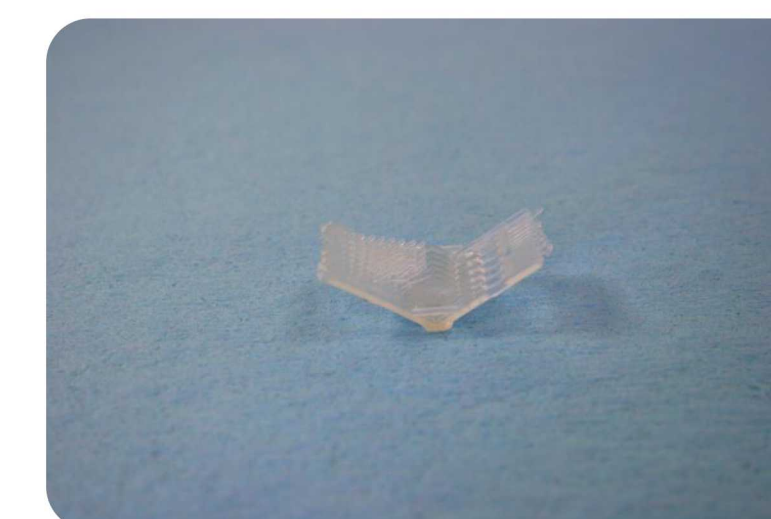
The prototype shown to the right includes:

- UV backlit build plate
- Stretched film resin tray
- Modified support plate made of acrylic.
- High power, high resolution (2560x1600) UV projector



### Standard Resin

- Genesis (By Tethlon 3D)
- 25  $\mu$ m layer resolution
- Scratched copper build plate
- *Basic functionality proven*



### Silica Loaded

- Porcelite (By Tethlon 3D)
- 25  $\mu$ m layer resolution
- UV backlit build plate
- *Ability to build using loaded resins proven*



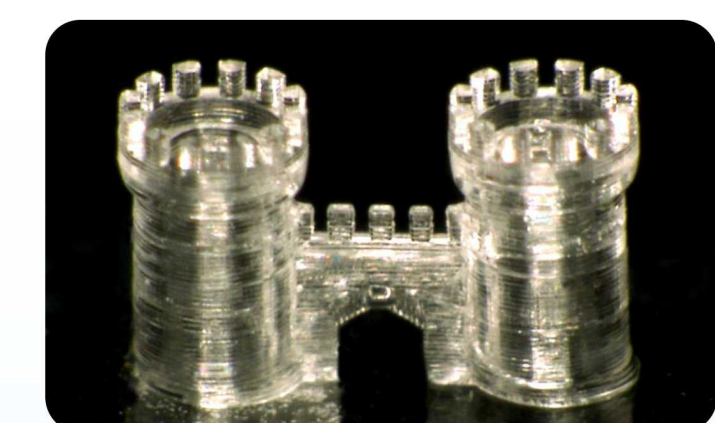
### Alumina Loaded

- Alumina loaded resin (Custom material)
- 25  $\mu$ m layer resolution
- UV backlit build plate
- *Proven some functionality building with alumina*



## Future Work

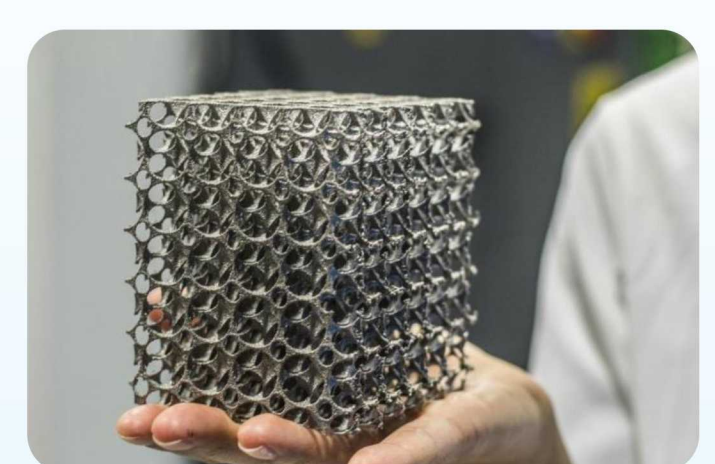
- Finalize prototype design and further iterations
- Custom control program
- Larger UV-backlit build plate, for larger production of printed parts without creating longer print times (when using DLP technology)
- Heated build chamber, for finer resolution by exploiting changes in resin curing at different temperatures (Hot Lithography).
- Alternative high-performance materials such as glass and metals



<https://phys.org/news/2017-04-d-glass.html>



<http://tethlon3d.com/product/porcelain-ceramic-resin/>



<http://engatech.com/metal-3d-printing/>





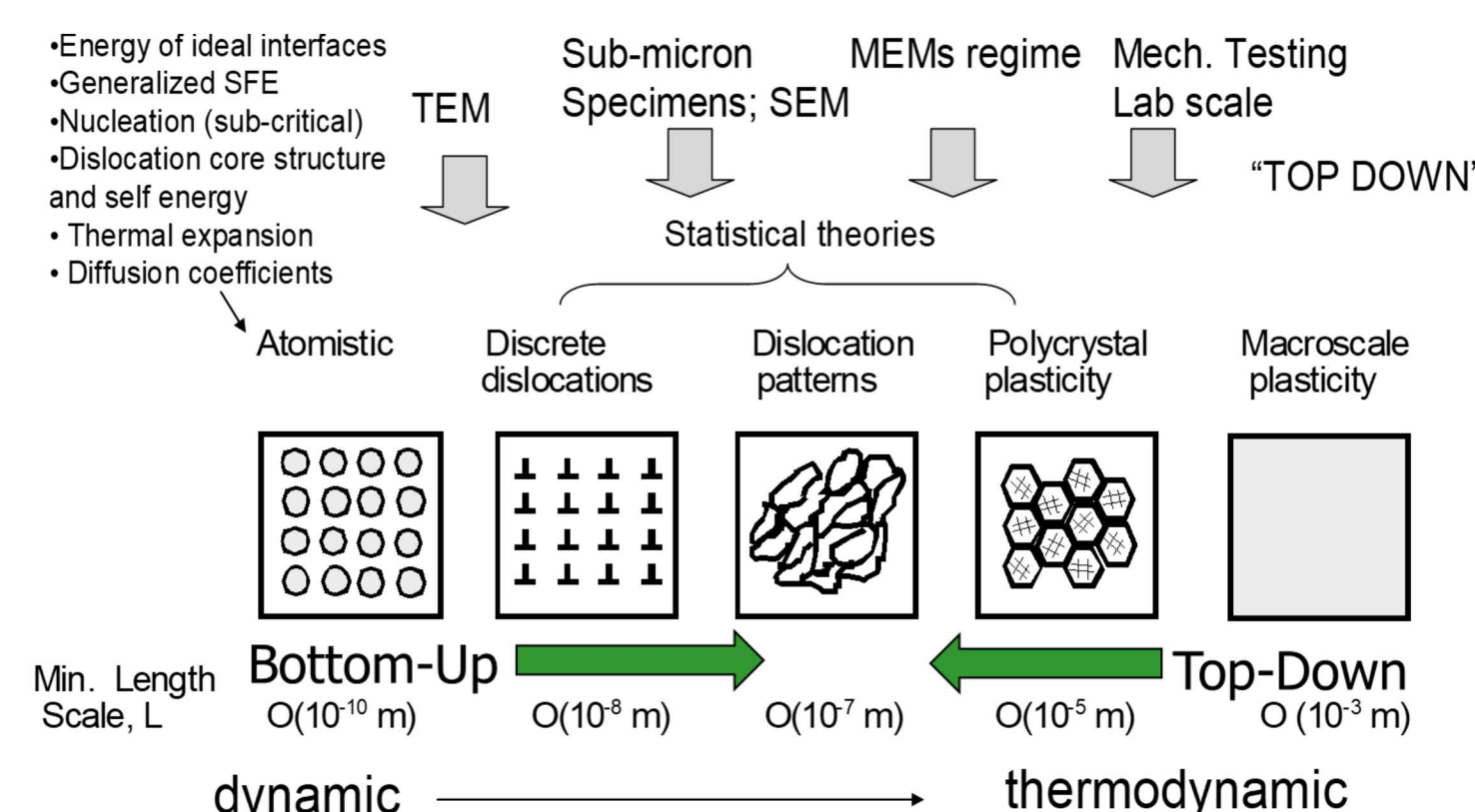
## Introduction / Motivation

Computational materials models exist at multiple scales with different parameterizations.

- How do models/scales connect?
- Which data are relevant to a prediction?

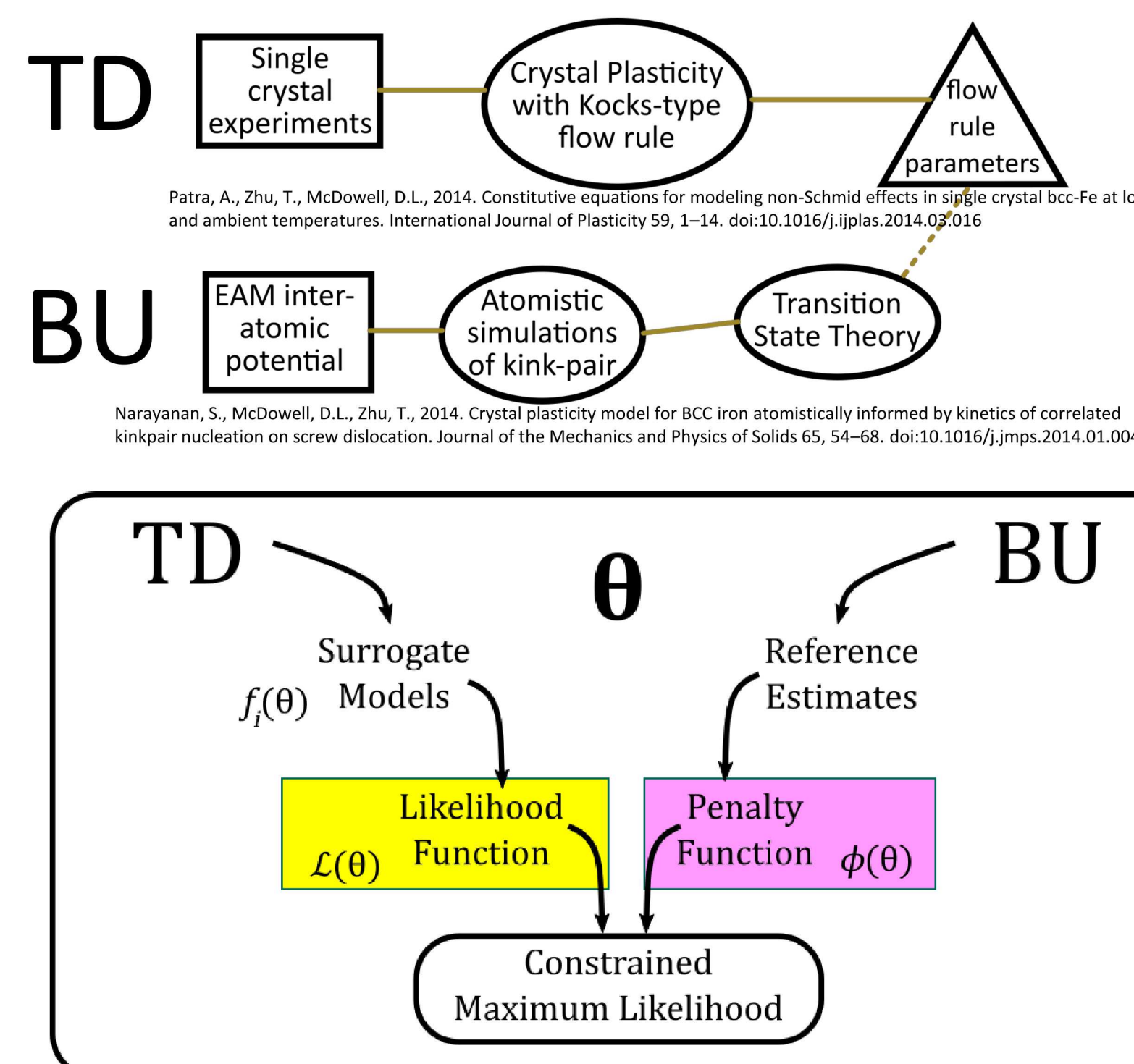
**GOAL:** Advance data collection strategies to:

- Perform Top-Down/Bottom-Up reconciliation of parameters
- Test empirical support of a connection between models



## Approach

### Reconciled Top-down and Bottom-up Hierarchical Multiscale Calibration of bcc Fe Crystal Plasticity



- Calibration parameters:  $\theta$
- To be TD and BU, models must meet in the middle  $\rightarrow$  value(s) for  $\theta, \hat{\theta}_{\text{TDBU}}$

- Where is the middle?
- Use likelihood-based methods to reconcile
- Constrained Likelihood: Maximum  $\rightarrow$  middle

Top-Down:

$$\text{Likelihood}(\theta) \propto \exp \left( -\frac{1}{2} \sum_{i=1}^{N_{\text{data}}} \frac{(f_i(\theta) - Y^E(x_i))^2}{\sigma_{\text{exp},i}^2} \right)$$

Bottom-Up:

$$\text{Penalty}(\theta)_j = \exp \left( -\frac{1}{2} \frac{(\theta_j - \hat{\theta}_{\text{BU},j}^{\text{ref}})^2}{\sigma_p^2} \right)$$

$$\text{Constrained Likelihood}(\theta) = \text{Likelihood}(\theta) \prod_{j=1}^{n_p} \text{Penalty}(\theta)_j$$

$$\text{Reconciliation} \rightarrow \max(\text{CLF}(\theta)) \Rightarrow \hat{\theta}_{\text{TDBU}}$$

## Challenges

Need to evaluate the difference between  $\hat{\theta}_{\text{TDBU}}$  and  $\hat{\theta}_{\text{TD}}$ . What is the cost and uncertainty associated with adding bottom-up information?

We formulate a connection cost “u” between TD and BU routes:

$$\text{CLF}(\theta | \text{Uncertainty}) = \text{CLF}(\theta)^u, u \in (0,1]$$

$u = 0$ : infinite connection cost, no agreement at all between TD and BU data

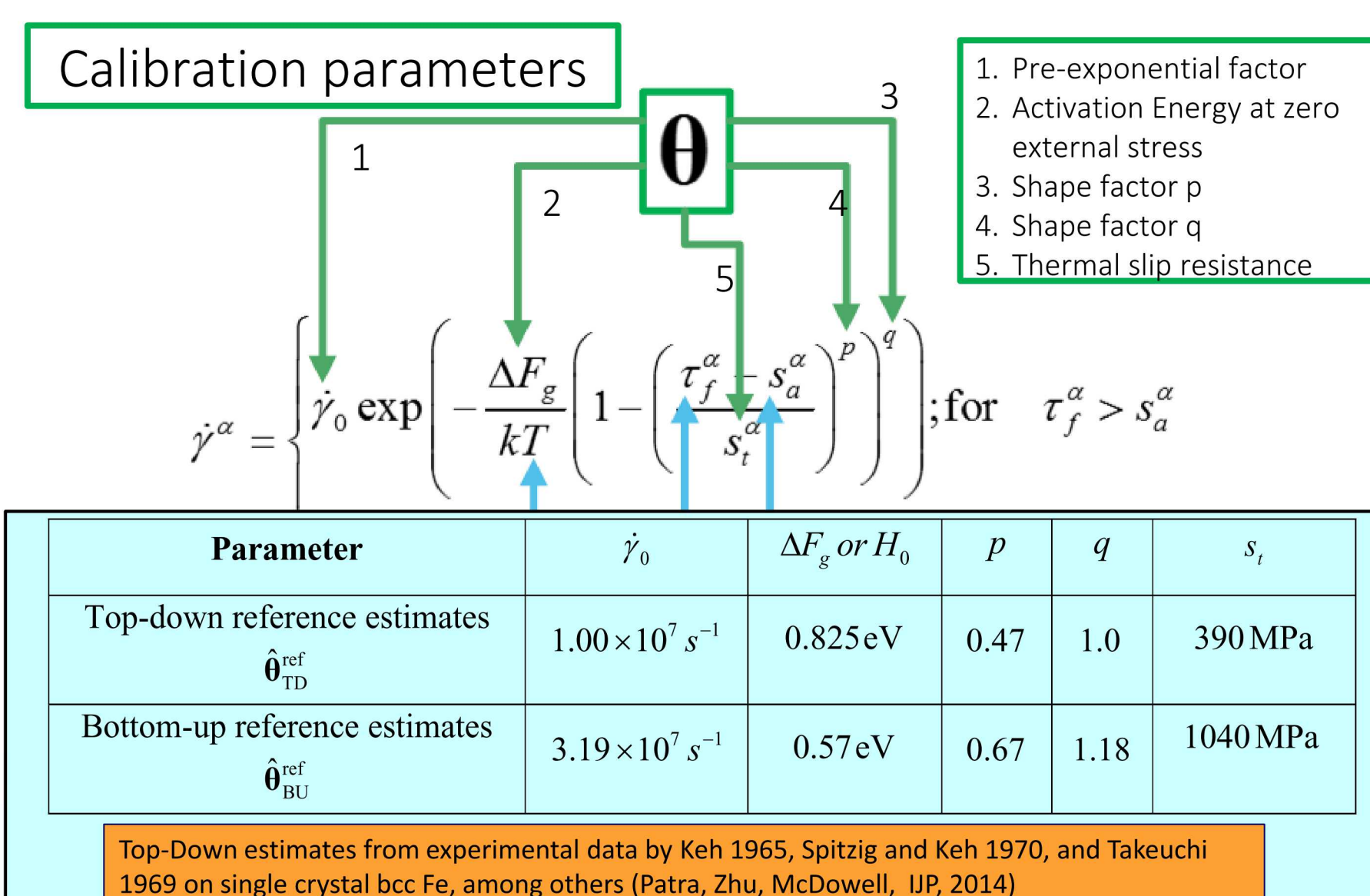
$u = 1$ : no connection cost; no disagreement between TD and BU data

$$u = \frac{1}{\max\{\text{SSE} + \text{Penalty}(\hat{\theta}_{\text{TDBU}}), 1\}} = \frac{1}{\max\{-2 \ln(\text{CLF}(\theta | \theta = \hat{\theta}_{\text{TDBU}})), 1\}}$$

$$\text{PDF}(\theta) = \frac{\text{CLF}(\theta)^u \text{Prior}(\theta)}{\text{Normalizing Const.}}$$

$$\sigma_\theta^2 = \int_{\theta\text{-space}} \text{PDF}(\theta) \sum_{j=1}^{n_p} (\theta_j - E[\theta_j])^2 d\theta$$

## Results



| $\hat{\theta}_{\text{BU}}^{\text{ref}}$ | $\dot{\gamma}_0$                  | $\Delta F_g \text{ or } H_0$ | p    | q    | $s_i$    |
|---|-----------------------------------|------------------------------|------|------|----------|
| Simulated                               | $3.19 \times 10^7 \text{ s}^{-1}$ | 0.57 eV                      | 0.67 | 1.18 | 1040 MPa |
| “cooperative”                           | $1.58 \times 10^7 \text{ s}^{-1}$ | 0.75 eV                      | 0.48 | 1.3  | 540 MPa  |
| “spurious”                              | $5.00 \times 10^5 \text{ s}^{-1}$ | 0.95 eV                      | 0.8  | 1.5  | 1100 MPa |

| Top-down reference estimates $\hat{\theta}_{\text{TD}}^{\text{ref}}$ | $\dot{\gamma}_0$                  | $\Delta F_g \text{ or } H_0$ | p    | q   | $s_i$   |
|--|-----------------------------------|------------------------------|------|-----|---------|
|  | $1.00 \times 10^7 \text{ s}^{-1}$ | 0.825 eV                     | 0.47 | 1.0 | 390 MPa |

| BU estimate $\hat{\theta}_{\text{BU}}^{\text{ref}}$ | Calib. | SSE  | SSE + Penalty | u                     | $\sigma_\theta^2$ |
|---|--------|------|---------------|-----------------------|-------------------|
| Simulated   | TD     | 37.7 | 161.7         | $5.98 \times 10^{-2}$ | <b>0.296</b>      |
|   | TDBU   | 83.1 | 102.3         | $1.17 \times 10^{-2}$ | <b>0.317</b>      |
| “cooperative”                                       | TD     | 37.7 | 81.0          | $5.98 \times 10^{-2}$ | <b>0.296</b>      |
|   | TDBU   | 56.1 | 57.5          | $4.34 \times 10^{-2}$ | <b>0.249</b>      |
| “spurious”  | TD     | 37.7 | 262.6         | $5.98 \times 10^{-2}$ | <b>0.296</b>      |
|   | TDBU   | 78.0 | 224.4         | $5.69 \times 10^{-3}$ | <b>0.340</b>      |

### Results summary

- u decreases with more data (implying more cost)
- After including the atomistic simulations, the  $\sigma_\theta^2$  value increased versus TD-only except for cooperative case
- Interpretation: only cooperative connection is helpful
- Diagnosis: many lines of inquiry to follow
- Parameter set may be a limitation, Peierls stress may be Temp dependent
- Mori, H., Temperature and Stress Dependence of Mobility of Screw Dislocation in BCC Iron, Solid State Phenomena, vol. 258, pp. 17–20, 2017.
- Choice of interatomic potential / constitutive laws may alter outcome
- Lim, H., Hale, L.M., Zimmerman, J.A., Battaile, C.C., and Weinberger, C.R., A multi-scale model of dislocation plasticity in  $\alpha$ -Fe: Incorporating temperature, strain rate and non-Schmid effects, International Journal of Plasticity, vol. 73, pp. 100–118, 2015

## Partnerships

### Collaborative efforts

Tallman, A.E., Swiler, L.P., Wang, Y., McDowell, D.L., 2017. Reconciled Top-down and Bottom-up Hierarchical Multiscale Calibration of bcc Fe Crystal Plasticity. JMC 15. <https://doi.org/10.1615/IntJMultCompEng.2017021859>

Special Issue Editors Yan Wang of Georgia Tech and Laura Swiler of Sandia: ASCE-ASME Journal of Risk and Uncertainty in Engineering Systems, Part B: Mechanical Engineering. Special Issue on Uncertainty Quantification in Multiscale System Design and Simulation. ASME J. Risk Uncertainty Part B. 2017; 4(1):010301-010301-2. doi: 10.1115/1.4037447.

Laura Swiler served on Aaron Tallman’s thesis committee (Ph.D. defense in July 2018). Book chapter and papers coming.

Note: The figures have been reproduced with permission from (Tallman, 2017) and Tallman’s thesis: “Hierarchical Multiscale Materials Modeling: Calibration, Uncertainty Quantification, and Decision Support,” 2018.