

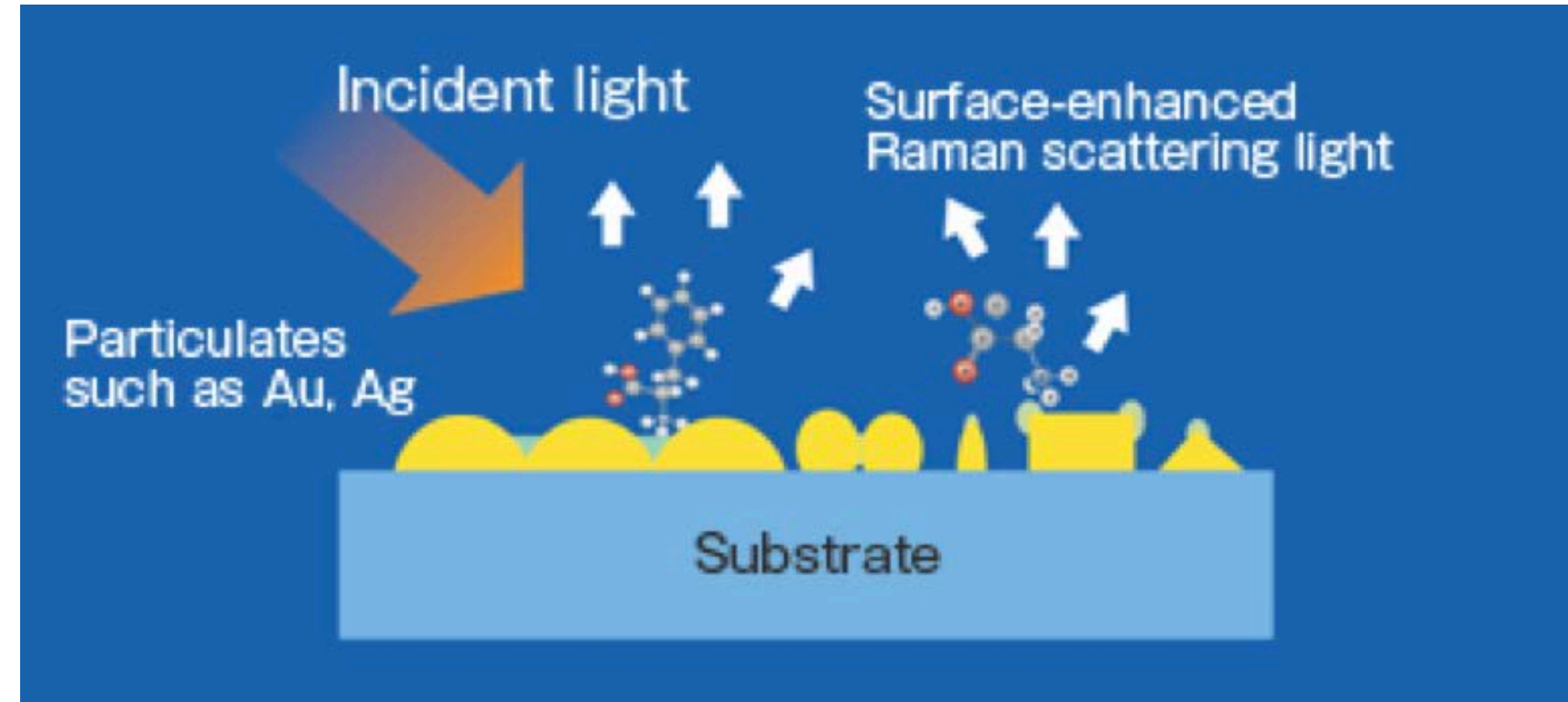
# Robust Design and Optimization of Nanoparticles: Morphology and Arrays

Y. He<sup>1,2</sup>, J. M. Razi<sup>5</sup>, R. Wang<sup>3</sup>, C. Forestiere<sup>4</sup>, L. Dal Negro<sup>3</sup>, R. M. Kirby<sup>5</sup>

1 University of North Texas, 2 New Mexico Tech, 3 Boston University, 4 Università degli Studi di Napoli Federico II, 5 University of Utah

## Research Area

Nanostructured metal can help to boost the performance of many plasmonic devices (e.g., Surface Enhanced Raman Scattering substrates, novel plasmonic nanosensors, etc.) by enhancing the near-field quantifies such as local electric field and optimizing the far-field synthetic parameters such as the scattering, absorption or extinction efficiencies.



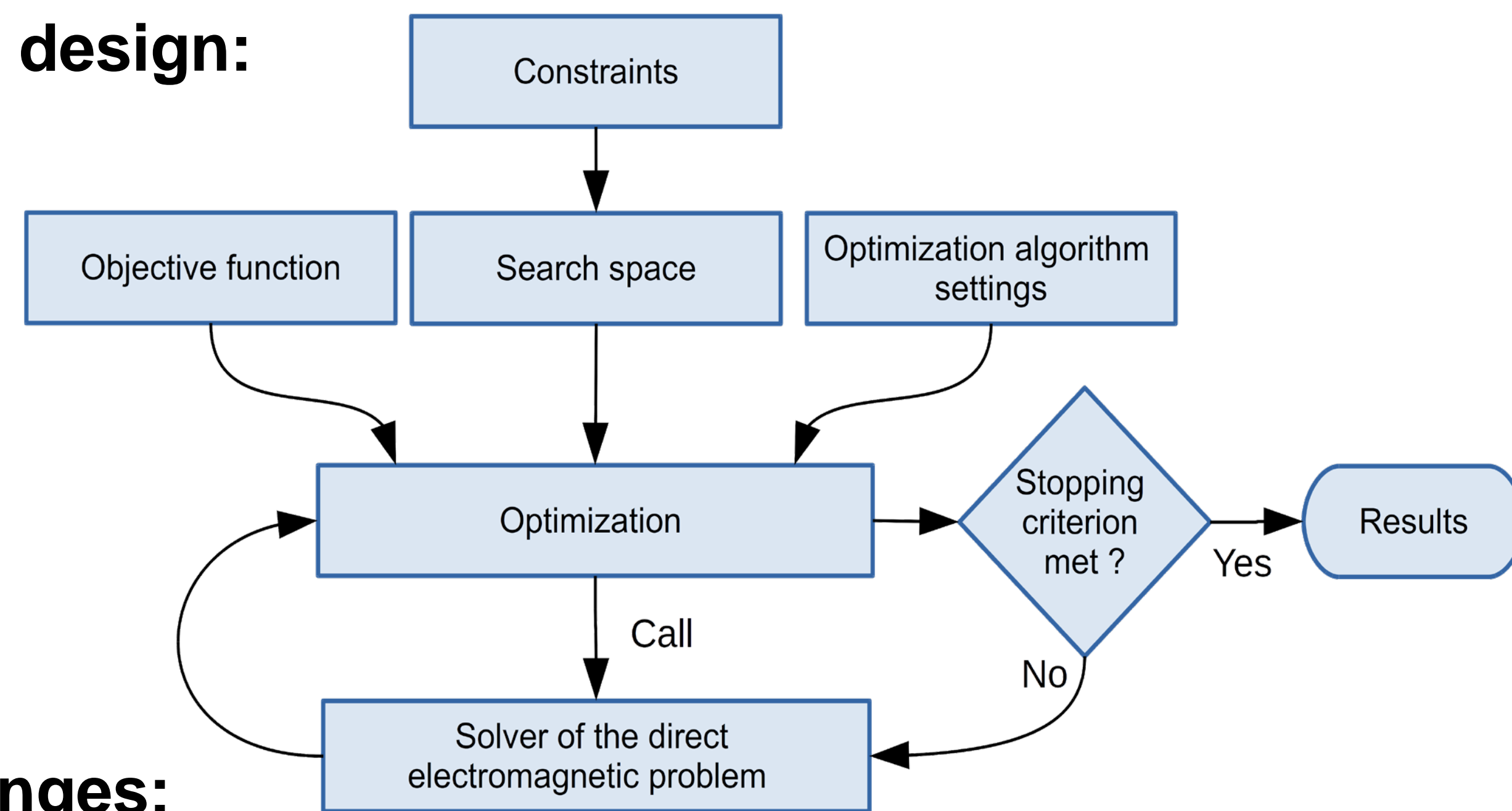
Uncertainty quantification (UQ) tools and efficient optimization algorithms are applied to the design of metallic nanoparticles.

## Goals

- UQ-guided robust design of individual nanoparticles for the optimization of near-field property (e.g., local field enhancement).
- Design of large-scale nanoparticle (Vogel Spiral) arrays for the optimization of far-field properties (e.g., absorption and scattering).

## Current State and Challenges

### Inverse design:

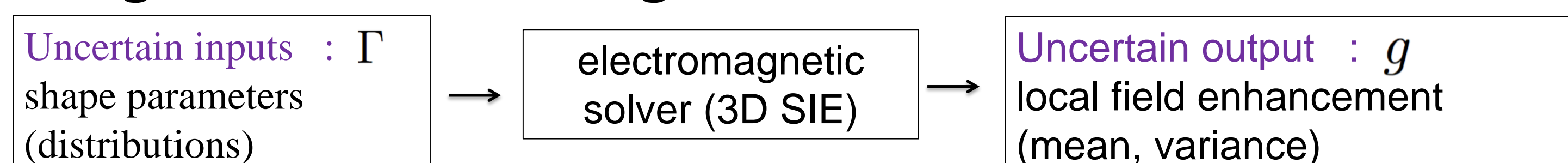


### Challenges:

- Direct inverse design may fail due to the uncertainty in the nanofabrication process. However, consideration of uncertainty in design process could be computationally expensive.
- Design of large-scale nanoparticle arrays requires an efficient optimization algorithm to deal with the large number of degrees of freedom.

## Strategy

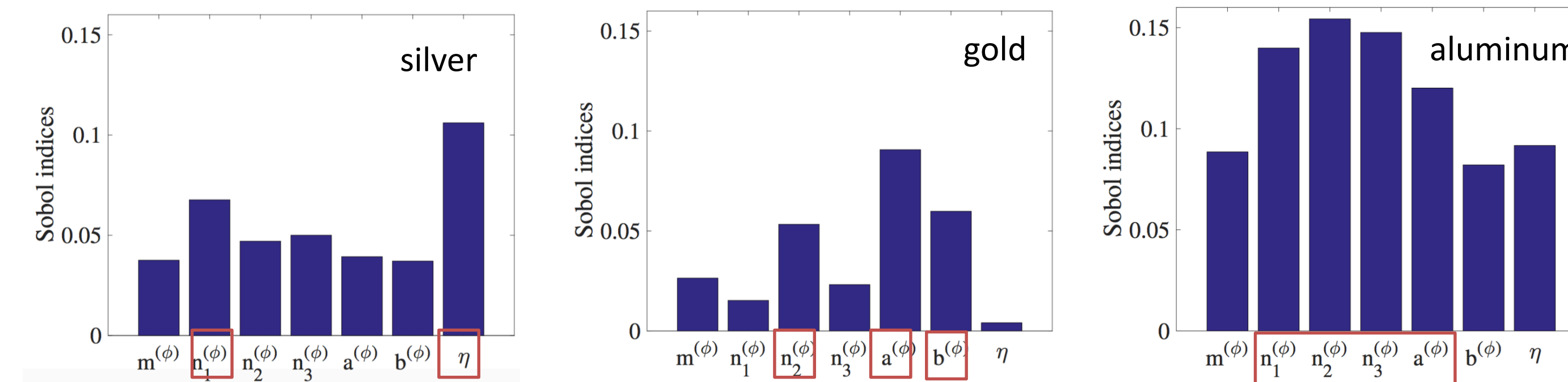
### UQ guided robust design:



## Strategy (contd.)

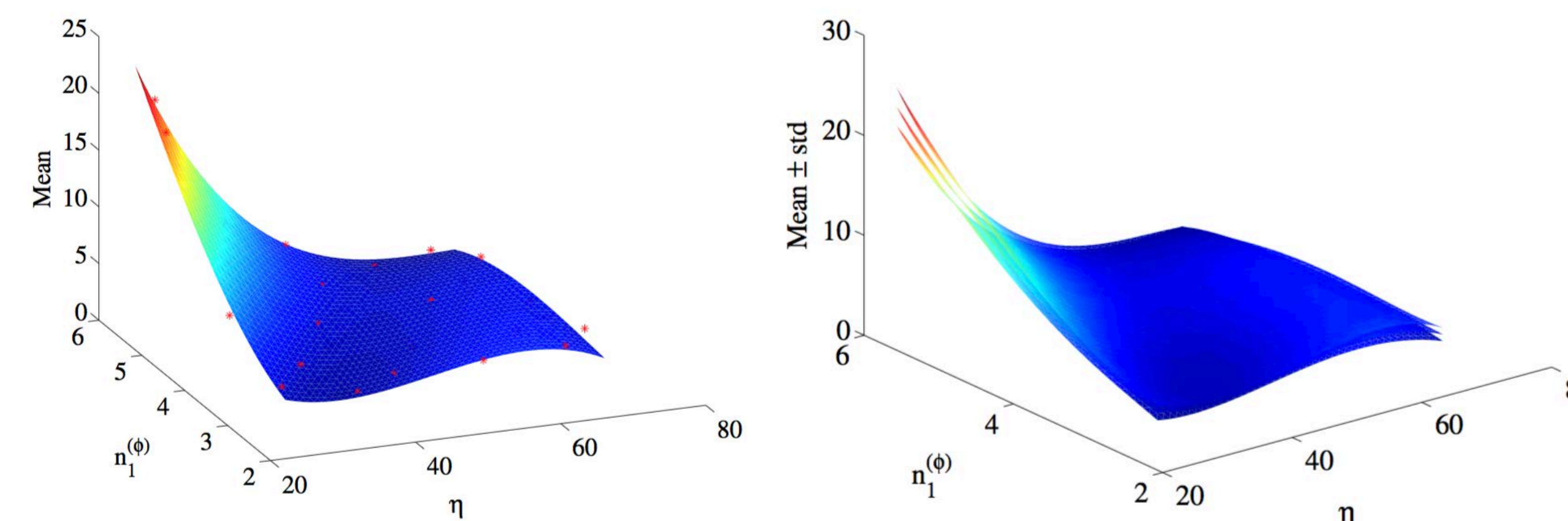
Robust design:  $\bar{\Gamma}^* = \operatorname{argmax}_{\bar{\Gamma}} E(g(\Gamma)) - \beta\sigma(g(\Gamma)) \quad \beta > 0$

- Global sensitivity analysis (identify “important” uncertain parameters)

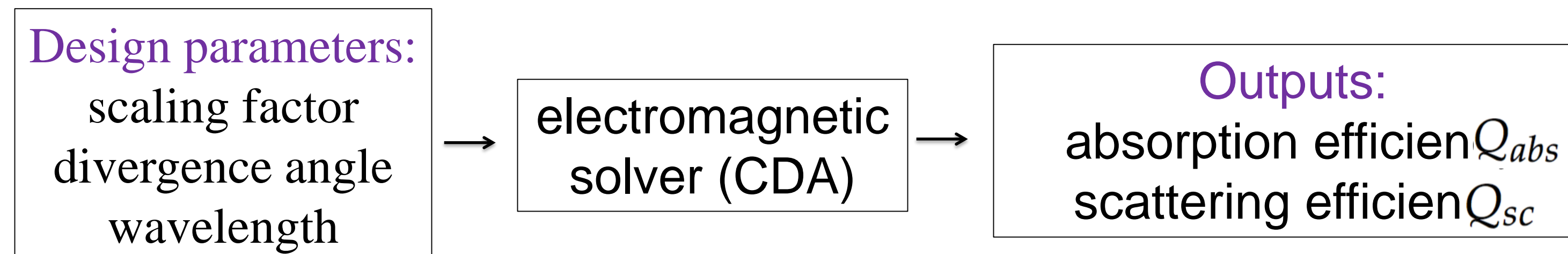


- Generalized polynomial chaos expansion

- Constructed over stochastic space of “important” parameters
- Used to approximate true mean and standard deviation



### Design of Vogel Spiral Arrays:



- Optimization algorithm (cyclic coordinate optimization)

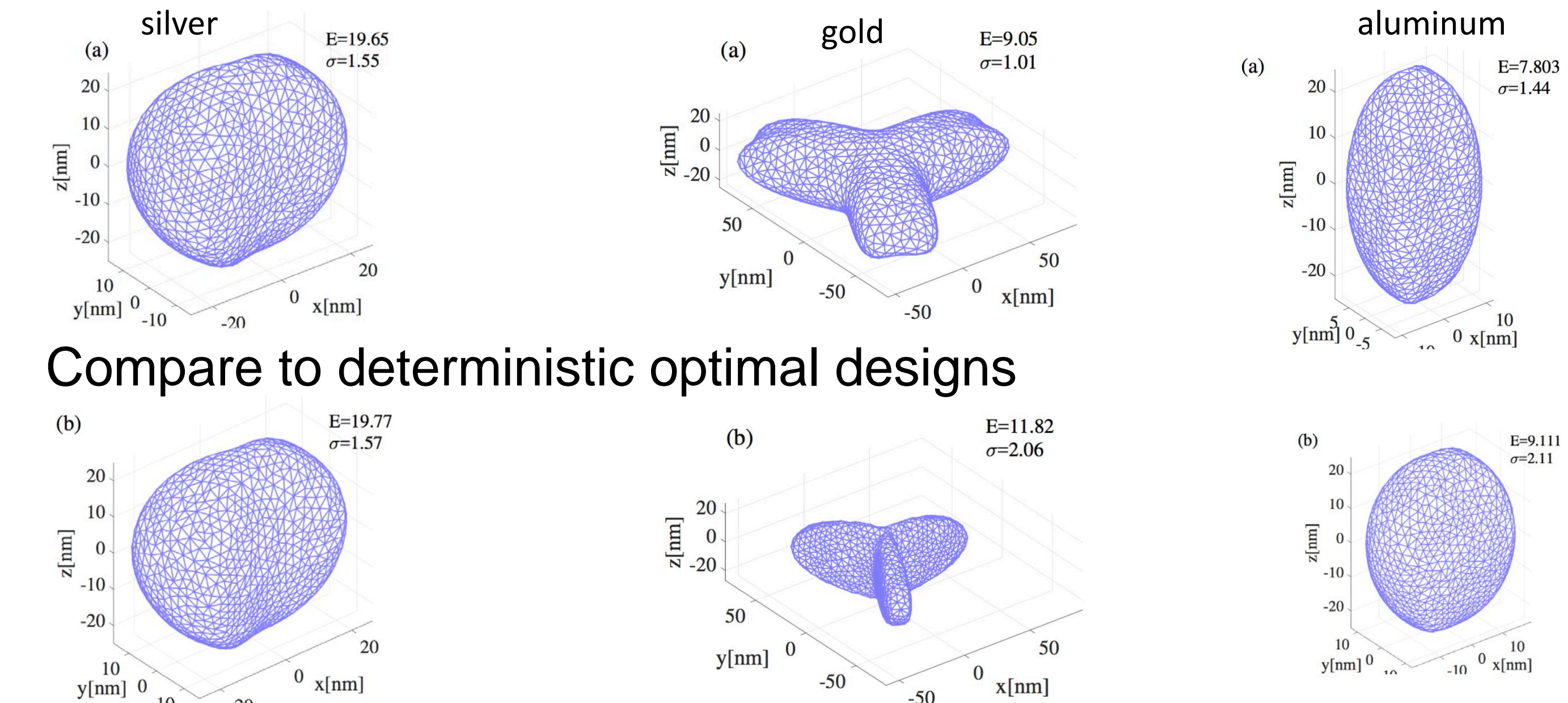
### Algorithm 1. cyclic coordinate optimization

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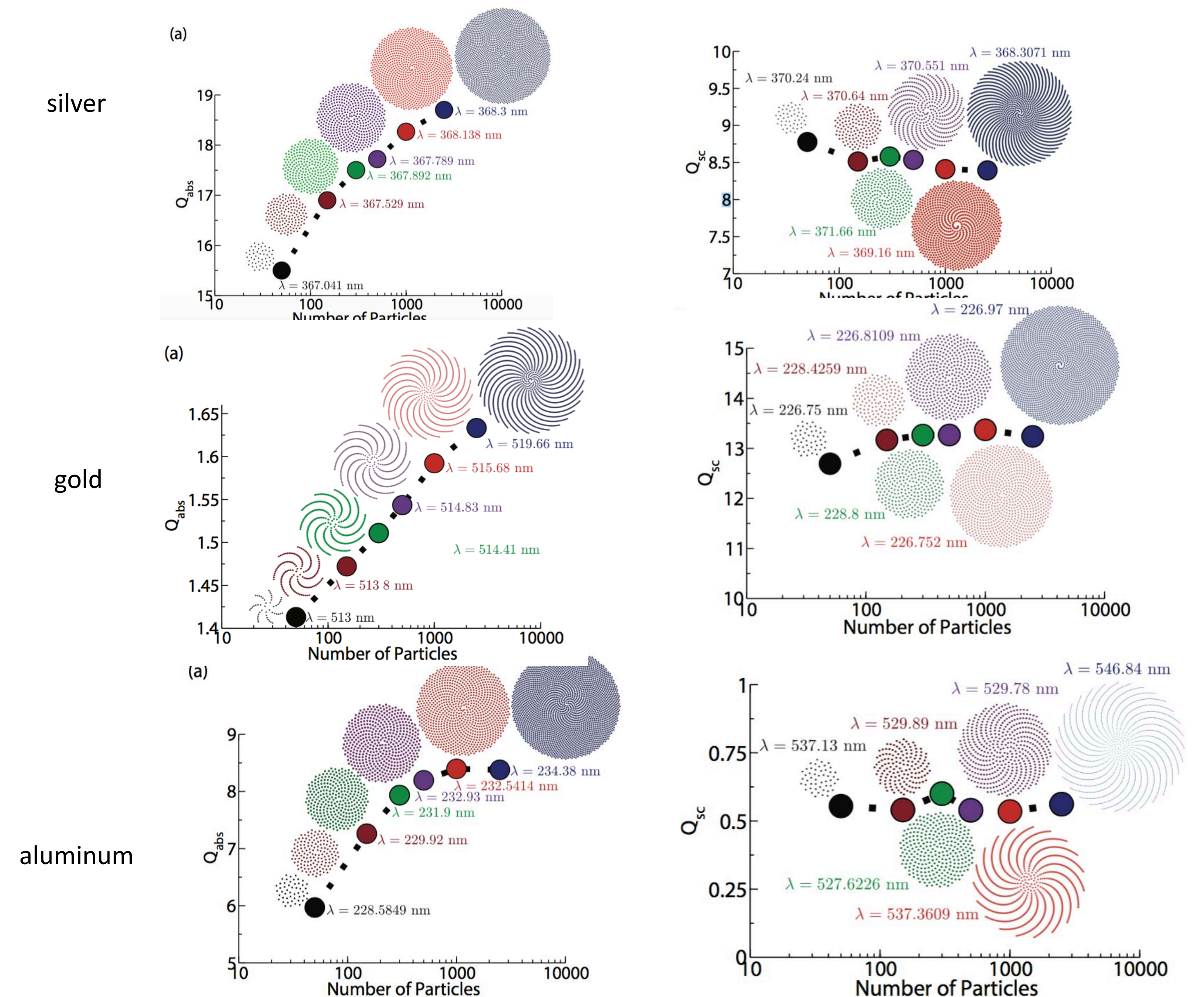
Set the number of particles N
Initialization of design variable X^(0) s.t. d ≥ 75
i ← 1
j ← 1
Compute objective function with initialized variables f^j
ε ← 10^3
while ε ≥ ε and a ≥ ε do
  while i ≤ 2N + 1 do
    x_i^j ← argmin_x f(x_1^j, ..., x_{i-1}^j, x, x_{i+1}^j, ..., x_{2N+1}^j)
    i ← i + 1
  Computing design parameters variation within a cycle: V ← X^new - X^old
  Pattern search: a ← argmin_a f(X^new + aV)
  ε ← |f^j - f^{j-1}|
  j ← j + 1
  
```

## Significant Accomplishments

- Robust design of individual particles (i.e., more robust)



- Optimal designs of large-scale particles for absorption/scattering



## Future Directions

- Develop UQ techniques to quantify uncertainty in design process.
- Apply UQ techniques and optimization tools to material designs of interest to the Army.

## Publications

Y. He, M. Razi, C. Forestiere, L. Dal Negro and R.M. Kirby, "Uncertainty Quantification Guided Robust Design for Nanoparticles' Morphology", *Computer Methods in Applied Mechanics and Engineering*, Volume 336, pages 578-593, 2018.  
 Mani Razi, Ren Wang, Yanyan He, Robert M. Kirby and Luca Dal Negro, "Optimization of Large-Scale Vogel Spiral Arrays of Plasmonic Nanoparticles", *Plasmonics*, Volume 14, Issue 1, pages 253-261, 2019.  
 Mani Razi, Robert M. Kirby and Aki Narayan, "Fast Predictive Multi-fidelity Prediction with Models of Quantized Fidelity Levels", *Journal of Computational Physics*, Volume 376, pages 992-1008, 2019.