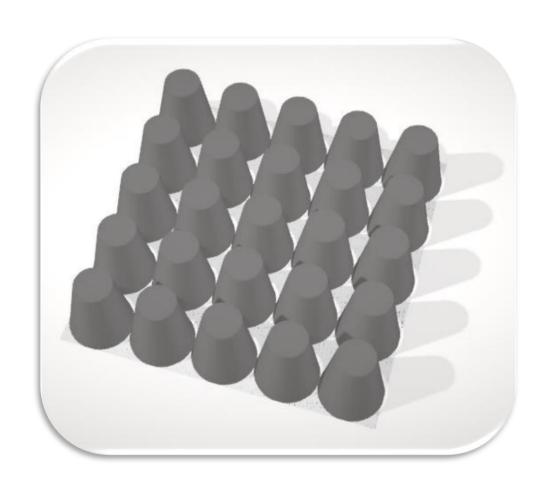


# **Use Case : Optimization of Anti-Reflective Moth-Eye Structures**

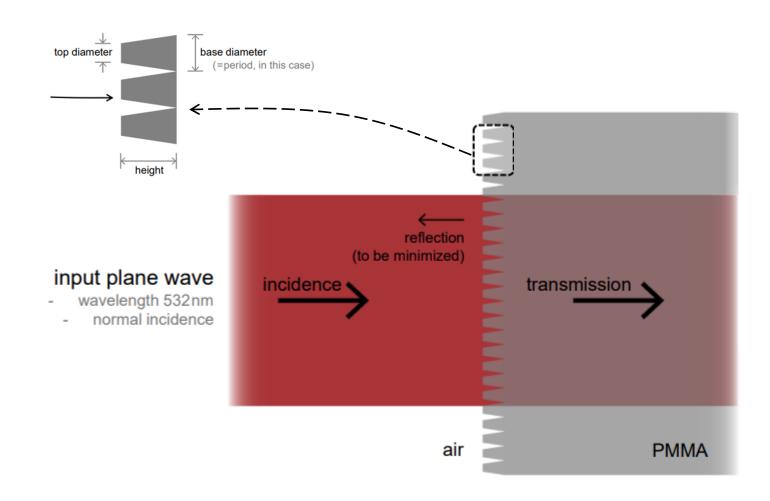


#### **Abstract**

Moth-eye structures are frequently utilized to impart antireflective (AR) properties to surfaces due to their unique and intricate designs. However, the complexity of these structures often leads to the presence of multiple local minima during optimization processes. To effectively address this challenge, two primary strategies can be employed. The first strategy involves conducting a rough parameter sweep to identify a promising starting point for subsequent local optimization. The second strategy entails initiating the optimization process with a global optimization algorithm.

In this use case, we will investigate and compare these two strategies in the context of an equidistant gridded moth-eye structure, evaluating their efficacy and outcomes in optimizing AR properties.

# **Optimization Task**



More Information Under:

Rigorous Analysis and Design of AR Moth-Eye

#### Parameter to be Optimized

> Reflection Efficiency

#### Varied Parameter

- Top Diameter (10~120nm)
- ➤ Height (50~500nm)

#### Algorithm

- Nelder-Mead
- Differential Evolutionary

In this case, we will attempt to minimize the reflection from the air-PMMA interface using two different optimization strategies: a local optimization algorithm (Nelder-Mead) and a global optimization algorithm (Differential Evolution).

#### **Overview of Results**

#### **/// Local Optimization**

Final Design #1

Height:141.47nm

Top Diameter:74.25nm

Overall Reflection Efficiency: < 0.0001%

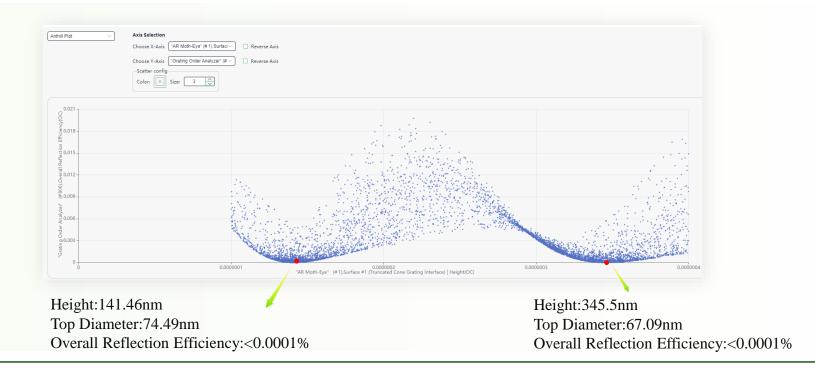
Final Design #2

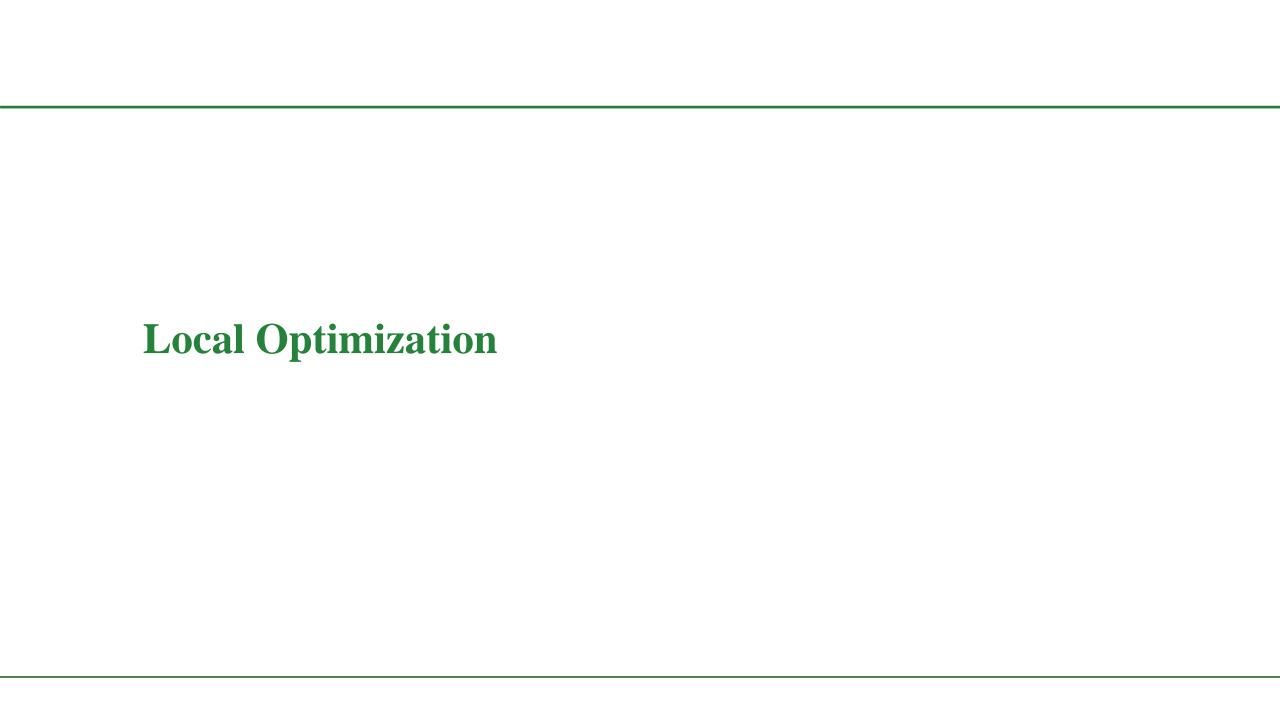
Height:346.0nm

Top Diameter:68.36nm

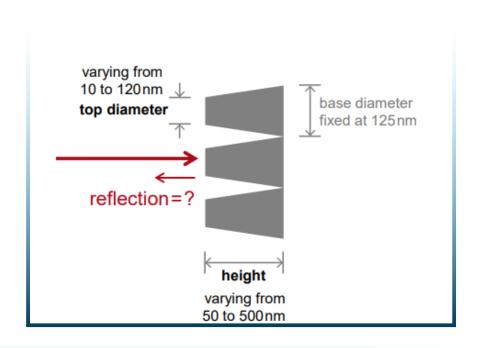
Overall Reflection Efficiency: < 0.0001%

#### **Global Optimization**

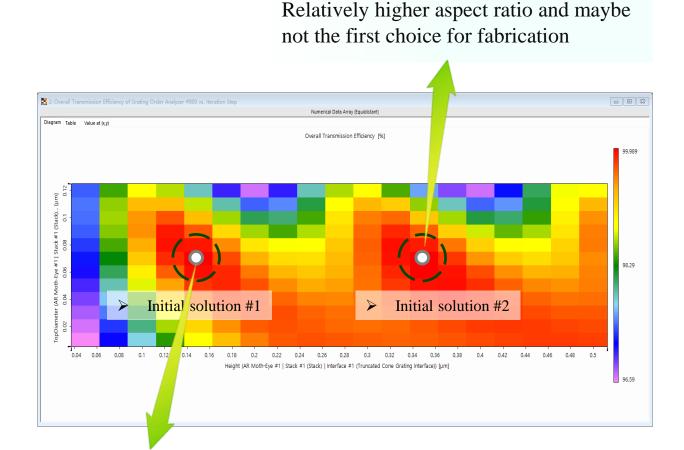




# Scanning over Parameter Space for Initial Solutions in VLF

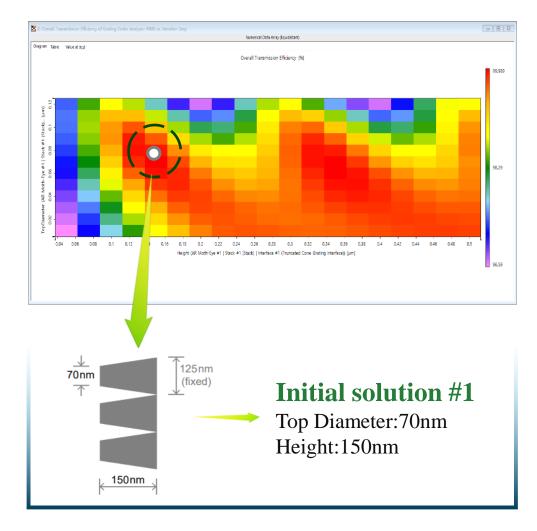


A parameter sweep is performed in order to find an adequate initial solution for the optimization. This is very useful for the upcoming local optimization algorithm, as local optimization is highly sensitive to the choice of starting point. A good starting point can significantly improve the optimization results.

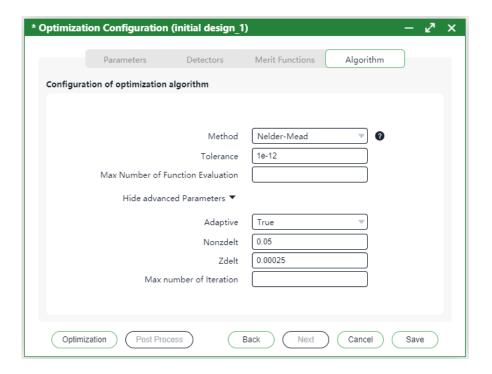


Relatively smaller aspect ratio and there fore preferable for fabrication

#### **Local Optimization-Initial Solutions #1**

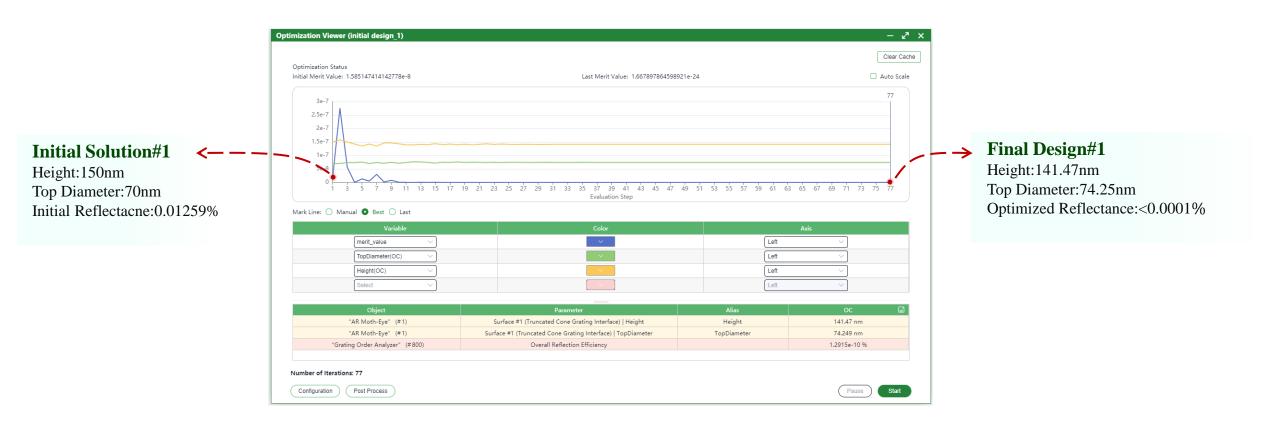


Regarding how to start and configure an optimization project please refer to: VirtualLab Fusion Optimization Package Tutorial

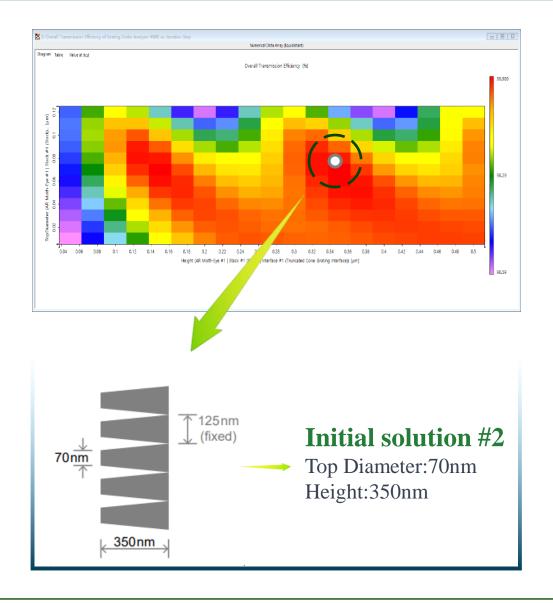


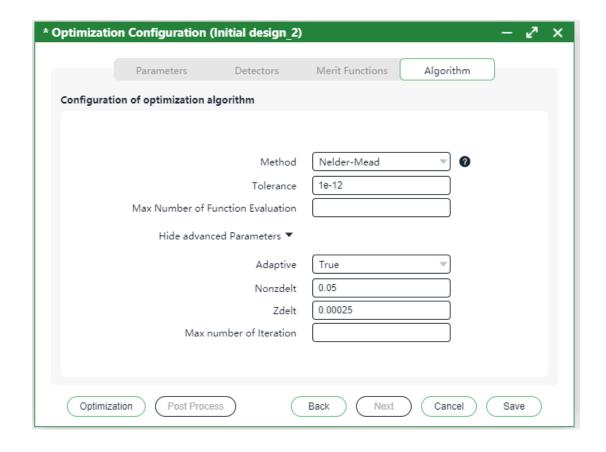
➤ We began the optimization process with initial solution #1 and employed the Nelder-Mead algorithm for optimization. Renowned as a widely-used local optimization algorithm, the Nelder-Mead algorithm is known for its robustness and broad applicability.

# Local Optimization-View the Result of Final Design#1



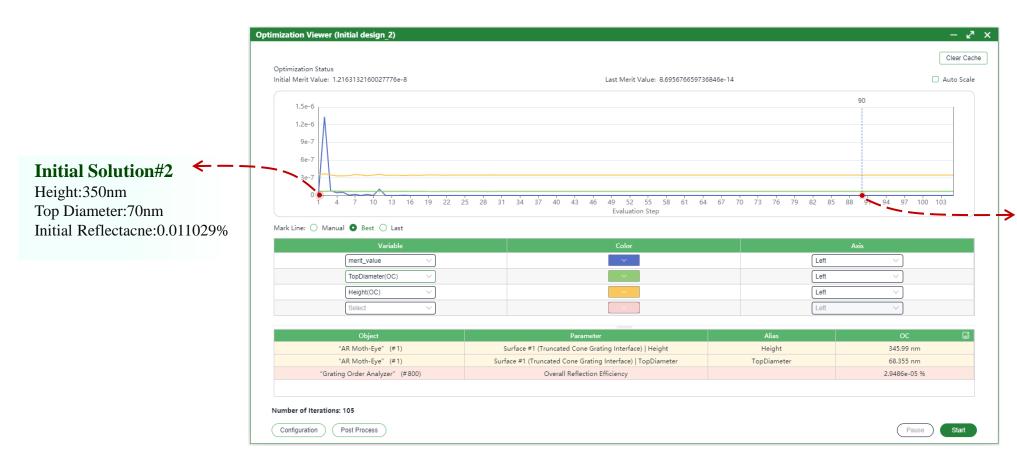
#### **Local Optimization-Initial Solutions #2**





➤ We began the optimization process with initial solution #2 and similarly employed the Nelder-Mead algorithm.

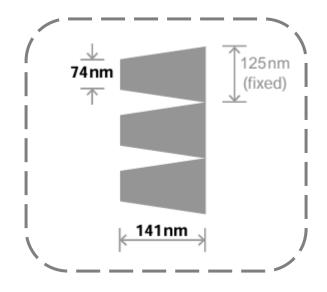
# **Local Optimization-View the Result of Final Design#2**



#### Final Design #2

Height:346.0nm
Top Diameter:68.36nm
Optimized Reflectance:0.0001%

# **Performance Analysis**

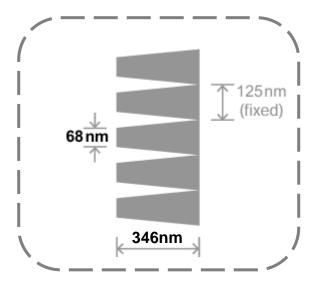


#### Final Design #1

Height:141.47nm

Top Diameter:74.25nm

Overall Reflection Efficiency: < 0.0001%



#### Final Design #2

Height:346.0nm

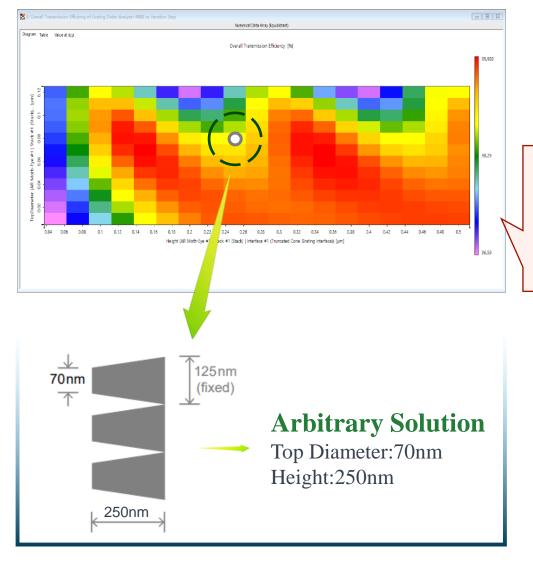
Top Diameter:68.36nm

Overall Reflection Efficiency: < 0.0001%

In the specified condition of normal incidence, although both designs can effectively reduce the reflectivity to less than 0.001%, we ultimately chose final design#1 because it has a smaller aspect ratio, making it more suitable for manufacturing.(Gratings with a high aspect ratio are more difficult to manufacture.).



# **Global Optimization-Arbitrary Solution**

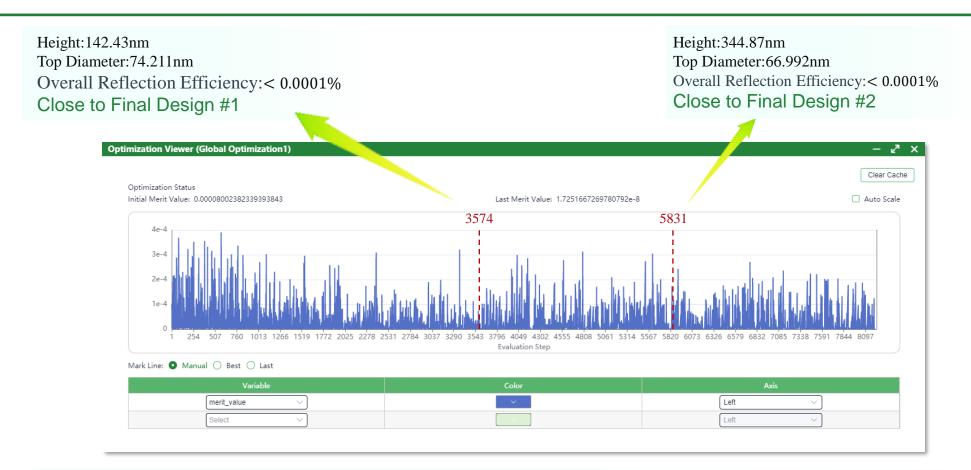


The parameter sweep is not required for global optimization; it is only used to demonstrate that the initial solution is not near a maximum

Optimization Configuration (Global Optimization) $- \kappa^{7} \times$			
Parameters	Detectors	Merit Functions Algorithm	
Configuration of optimization algorithm			
Method		Differential Evolutionary . 🔻	
Number of workers		50	
Max Iteration Number		1000	
Population Size		100	
Allow Dithering		True	
Mutation		0.7	
Recombination		0.8	
Relative Tolerance		0.000001	
Absolute Tolerance		0	
Show ad	vanced Parameters 🕨		
Optimization Pos	st Process Ba	ck Next Cancel Save	

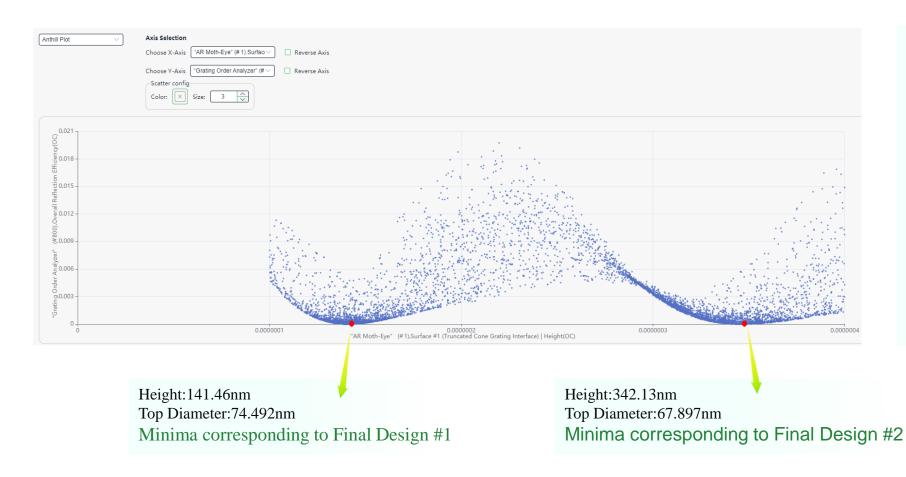
➤ We selected an arbitrary solution and used the differential evolution algorithm for optimization. The differential evolution algorithm is a global optimization algorithm with strong global search capabilities, fast convergence speed, and broad applicability.

# **Global Optimization-View Result**



By observing the optimization results, we can see that some values are close to Design 1, while others are close to Design 2 (as indicated by the red dashed lines in the figure). This suggests that two local optima can be found in the global optimization process. We will use post process to better illustrate this point.

# **Global Optimization-Post Process**



In the Anthill Plot, by setting Height as the x-axis and Reflectance as the y-axis, it can be observed that there are very low reflectance values at two different height positions. These two points correspond closely to the previous Final Design #1 and Final Design #2. This demonstrates the global optimization capability of the DE algorithm, as it found two local optima in the entire parameter space.

# **Document Information**

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category	Application Use Case	
further reading		