

An automated Bayesian optimization workflow for antimicrobial nanosurfaces

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SIBLEY GRADUATE RESEARCH SYMPOSIUM¹

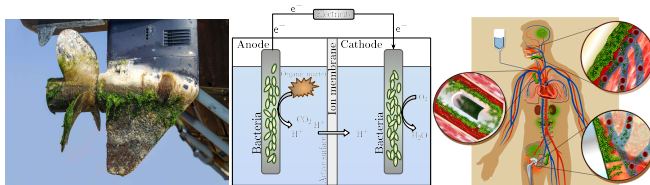
MAE, Cornell University

April 28, 2022

¹work supervised by Prof. Jingjie Yeo

General Background

Biofouling is a pain in ocean, energy engineering, biomedical treatments, etc. We hope to provide a **digital solution** for this hard problem.



Integrate



Materials 4.0



Design

Key Driver



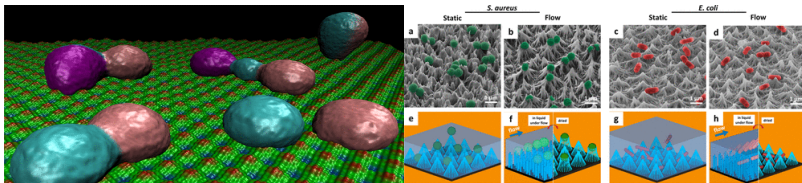
Digital twins + optimization

The recent "Materials 4.0 Initiative" drives huge innovations in materials-by-design with the "computational modeling + ML" paradigm in various fields.

¹<https://www.clubmarine.com.au/exploreboating/articles/32-3-Keeping-A-Clean-Bottom> ²<https://www.cs.montana.edu/webworks/projects/stevesbook/contents/chapters/chapter005/section002/blue/page005.html>

Inspirations & Technical Backgrounds

- The **super-hydrophobic surfaces** have been of interest for many years in the communities.
- Design the nanosurfaces (coatings) for antimicrobial properties have just recently been raised and the trend is growing drastically.
- **Bayesian optimization** for materials design is also becoming a new trend in the heated topic of ML algorithms for real-world problems.
- **Individual-based modeling**² is currently one of the most successful methods for accumulating M&M properties of biofilm modeling.

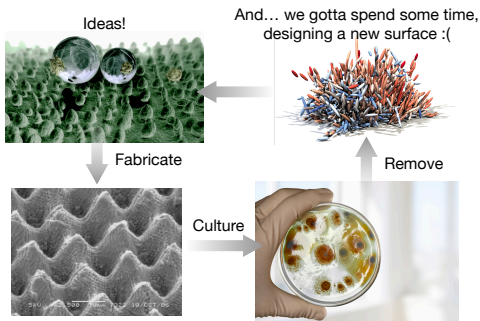


Zhang et al., *Langmuir*, 2019; Hizal et al., *ACS Appl. Mater. Interfaces*, 2017.

²https://hanfengzhai.net/file/Biofilm_review.pdf

When there's a problem, there's a solution!

Traditionally, how do we solve this problem?



If you rich and got plenty of time, you fine... :)) But can we solve this problem in a more "efficient" way?

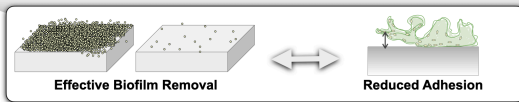
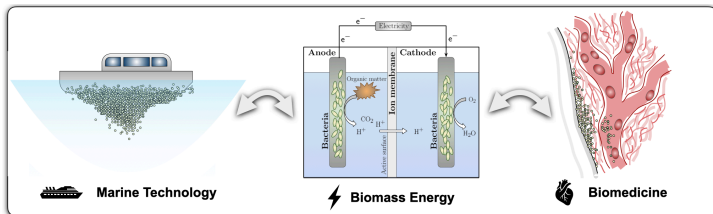
<https://www.youtube.com/watch?v=HF4blivJQ6o>

<https://phys.org/news/2007-01-lotus-leaf-imitated-plastic-femtosecond.html>

<https://researchoutreach.org/articles/breaking-down-fort-combatting-clinical-biofilms/>

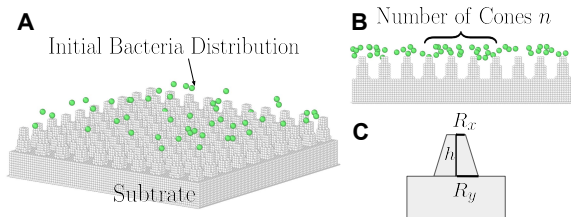
<https://www.ucl.ac.uk/~ucahppe/research.html>

Basic workflow

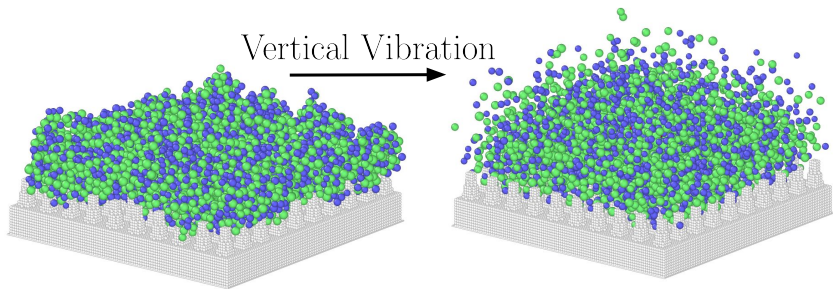
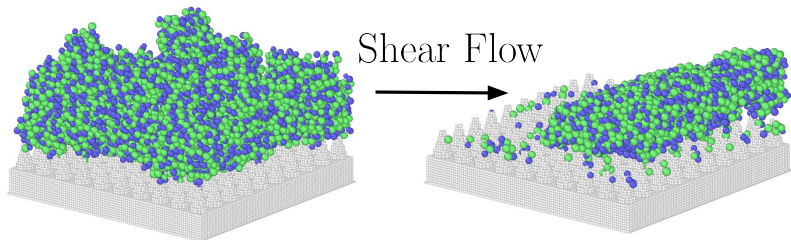


Numerical setup

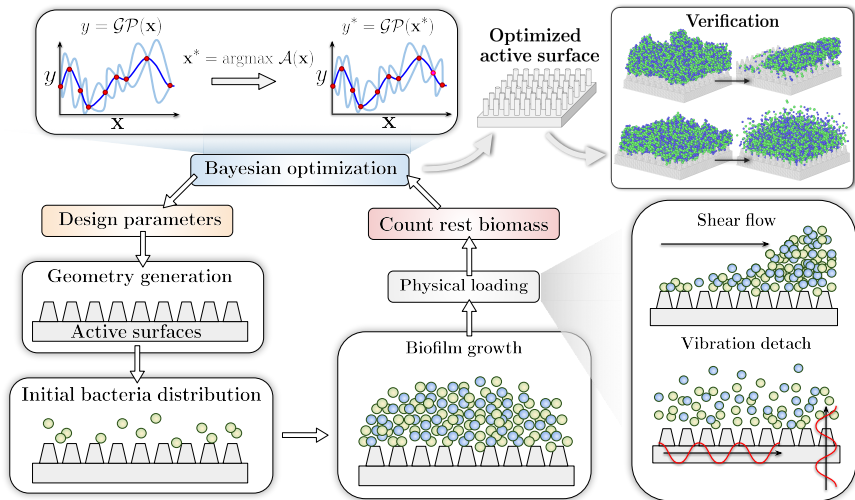
- **Geometric Design variables:** R_x, R_y, h, n
- **Simulation box:** Geometry: 4×10^{-3} m for x, y, z ; *Boundary conditions:* fixed BCs on x, y, z ; *Initial bacteria area:* 2×10^{-6} m
- **Bacteria cells:** *Initial No.:* 50; *Growth rate:* 0.00028; *Yield:* HET: 0.61 & EPS: 0.18. Monod growth model. HET: $K_s = 3.5 \times 10^{-5}$
- **Material properties:** *Heterotrophs:* $\rho = 150$; $d = 10^{-6}$; $m = 7.854 \times 10^{-17}$; *Substrate:* $\rho = 4410$; $d = 5^{-7}$; $m = 9.1875 \times 10^{-17}$.
- **Biofilm simulation:** *Grow:* $2 \times 10^4 (\times 10)$; *Shear:* $2 \times 10^4 (\times 2.5)$; *Vibration:* $1 \times 10^3 (\times 10^3)$;



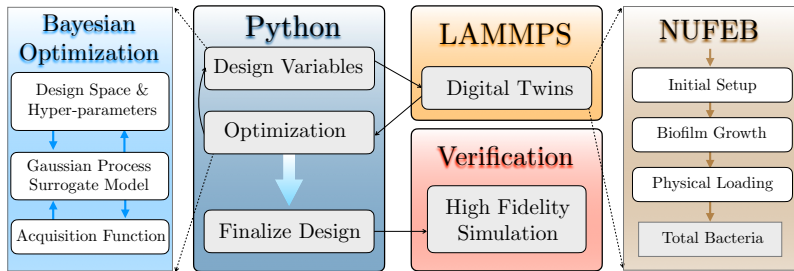
Biofilm removal: shear and vibration



Bayesian coupled workflow for materials design



Optimization workflow

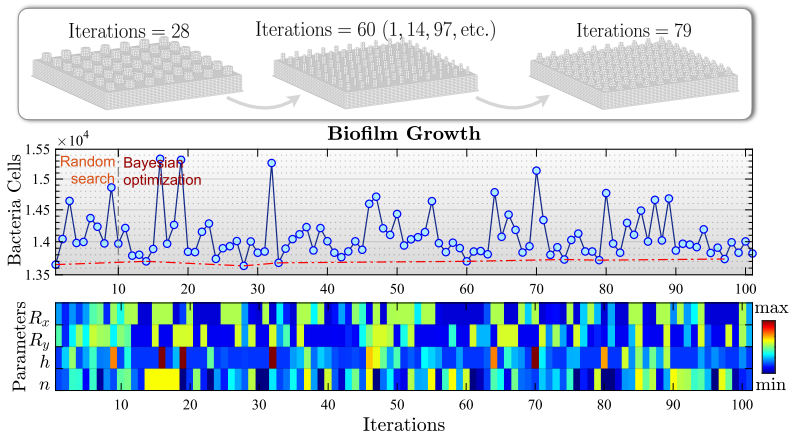


Technical Implementation

1. The whole optimization workflow is dependent on Python-LAMMPS interface.
2. Calculation on 100 CPU cores usually requires approximately 30 hours.
3. Variables are passed from randomization in Python to LAMMPS as a string (%s).

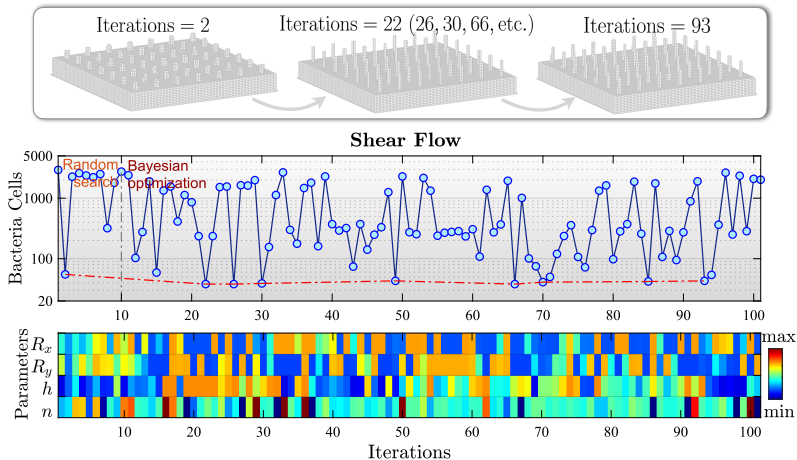
Results: pure biofilm growth

The active surfaces with shorter cones and mild thick shapes seem to effectively resist biofilm growth.



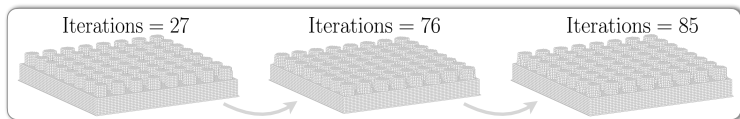
Results: shear flow detachment

For the shear flow biofilm detach, a "thin pillar" –shaped cone designs shows extraordinary biofilm removal effect.

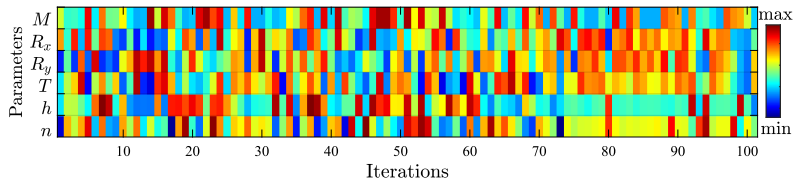
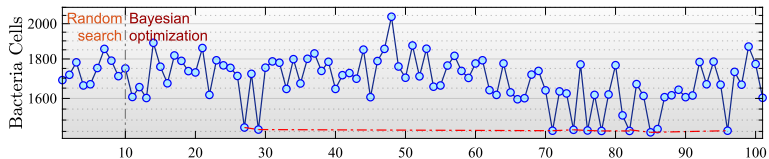


Results: vibration detachment

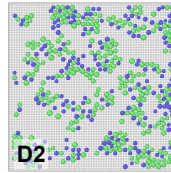
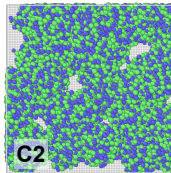
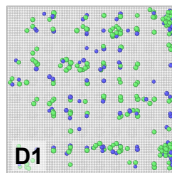
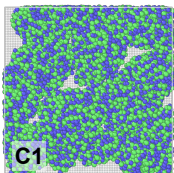
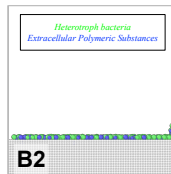
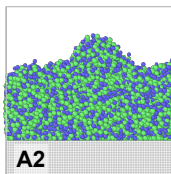
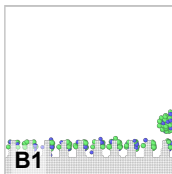
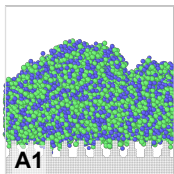
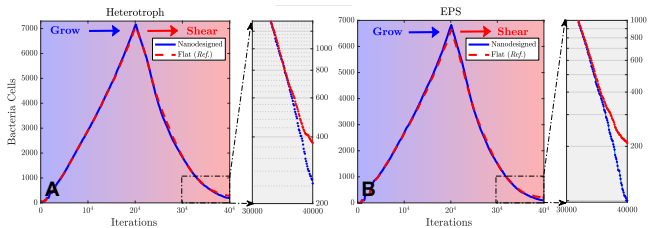
Strangely, but not strangely, for the vibration case, all the optimized active surfaces tend to exhibit very similar structures.



Vibration Detach



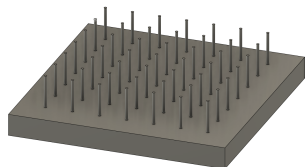
Verification



Summary & Discussions

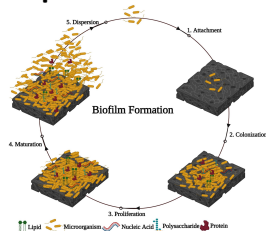
- This study presents a **digital solution** for materials design targeting antifouling problems with low cost and high efficiency.
- An automated **optimization workflow** enabled by discrete element multiphysics simulation is presented for materials design.
- Different **optimized geometries** based on different loading cases are generated from the workflow.

- **3D Printing**










Generated CAD file → 3D printing with hard materials for experimental verification

- **Experimental Validation**



Srinivasan et al., *Front. Microbiol.*, 2021

References

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-  Hizal et al., *ACS Appl. Mater. Interfaces.* 2017, 9, 13, 12118–12129.
-  Dongil Shin et al., *Adv Mat.* 2022, 34, 2106248.



Thanks for listening!

Any Questions...?