

A Whole Cell Biofilm Model

A simulation for antibacterial evaluation

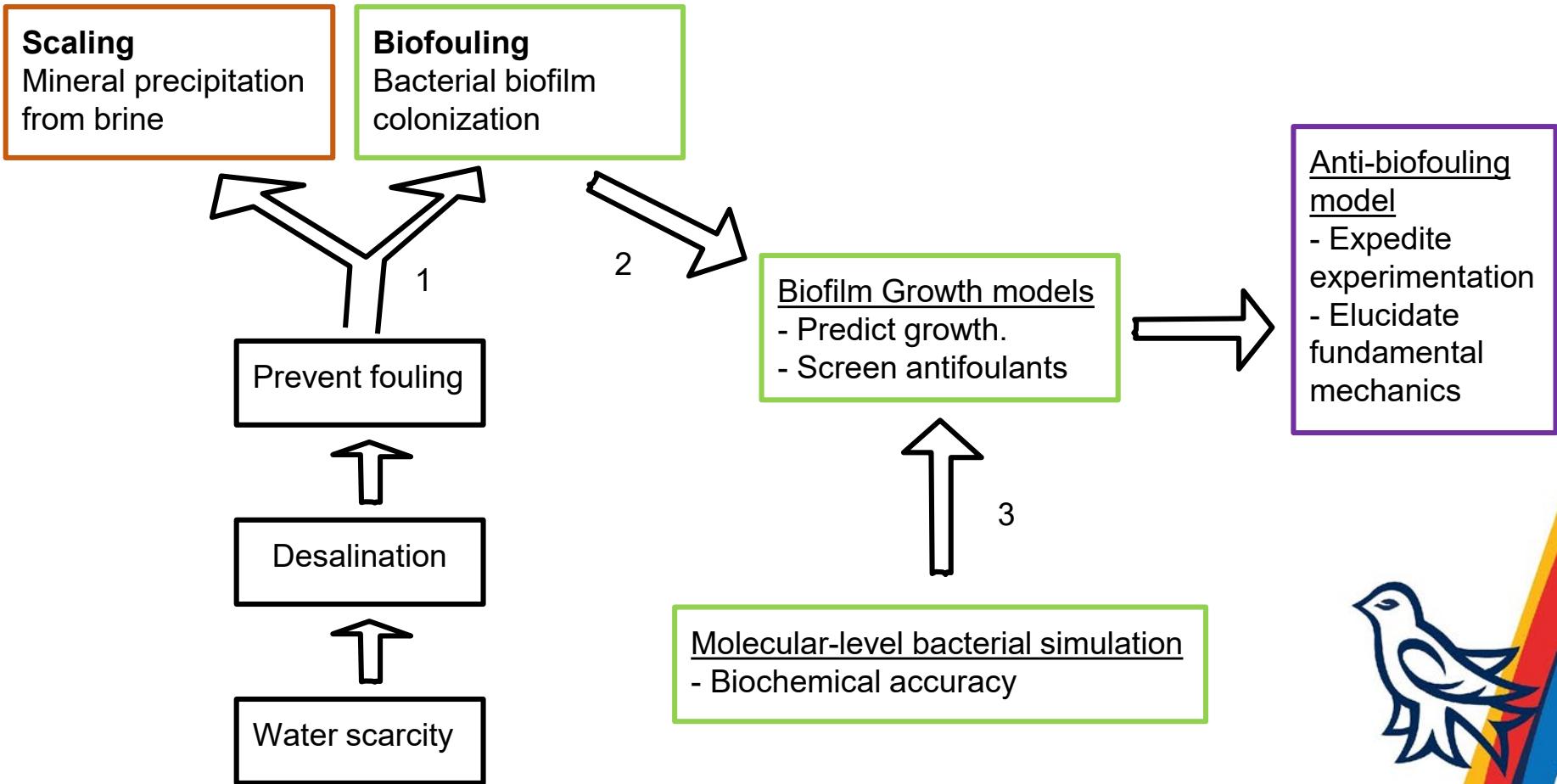
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ACS Spring National Meeting



University
of Victoria

Andrew Philip Freiburger

Desalination is hindered by biofouling



Experimental system

1) Can ${}^1\text{O}_2$ prevent biofilm formation?

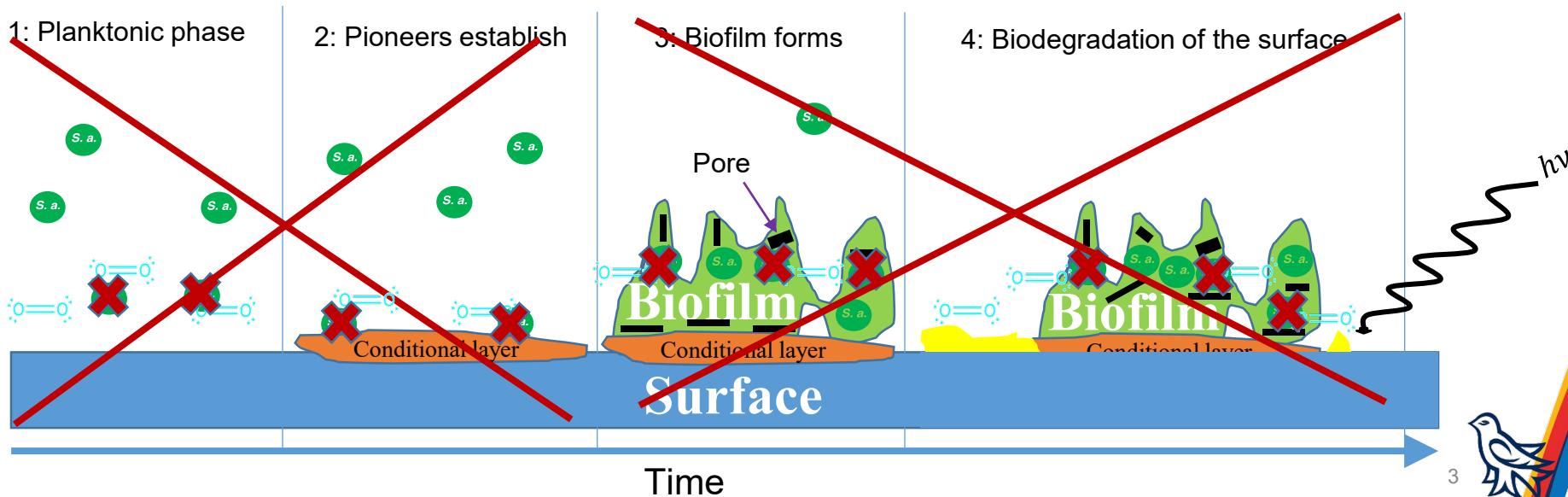
2) Can ${}^1\text{O}_2$ inactivate existing biofilms?

3) What are the requisite ${}^1\text{O}_2$ concentrations?

4) What mechanisms cause ${}^1\text{O}_2$ inactivation?

5) Is periodic dosing of ${}^1\text{O}_2$ sufficient for inactivation?

6) Does ${}^1\text{O}_2$ differentially affect (Gram +\-) species?



PDI oxidation – to scale

Staphylococcus aureus

($1.0 \mu\text{m}$ diameter
 50 nm membrane)

Bacterial dimensions:

S. aureus = sphere $[0.5, 1.5]\mu\text{m}$
membrane $[20, 80]\text{nm}$

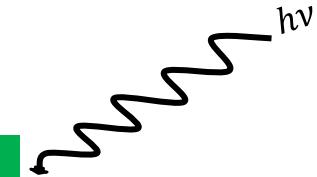
P. aeruginosa = rod $[0.5, 1]\mu\text{m} \times [1, 5]\mu\text{m}$

E. coli = rod $1\mu\text{m} \times 2\mu\text{m}$

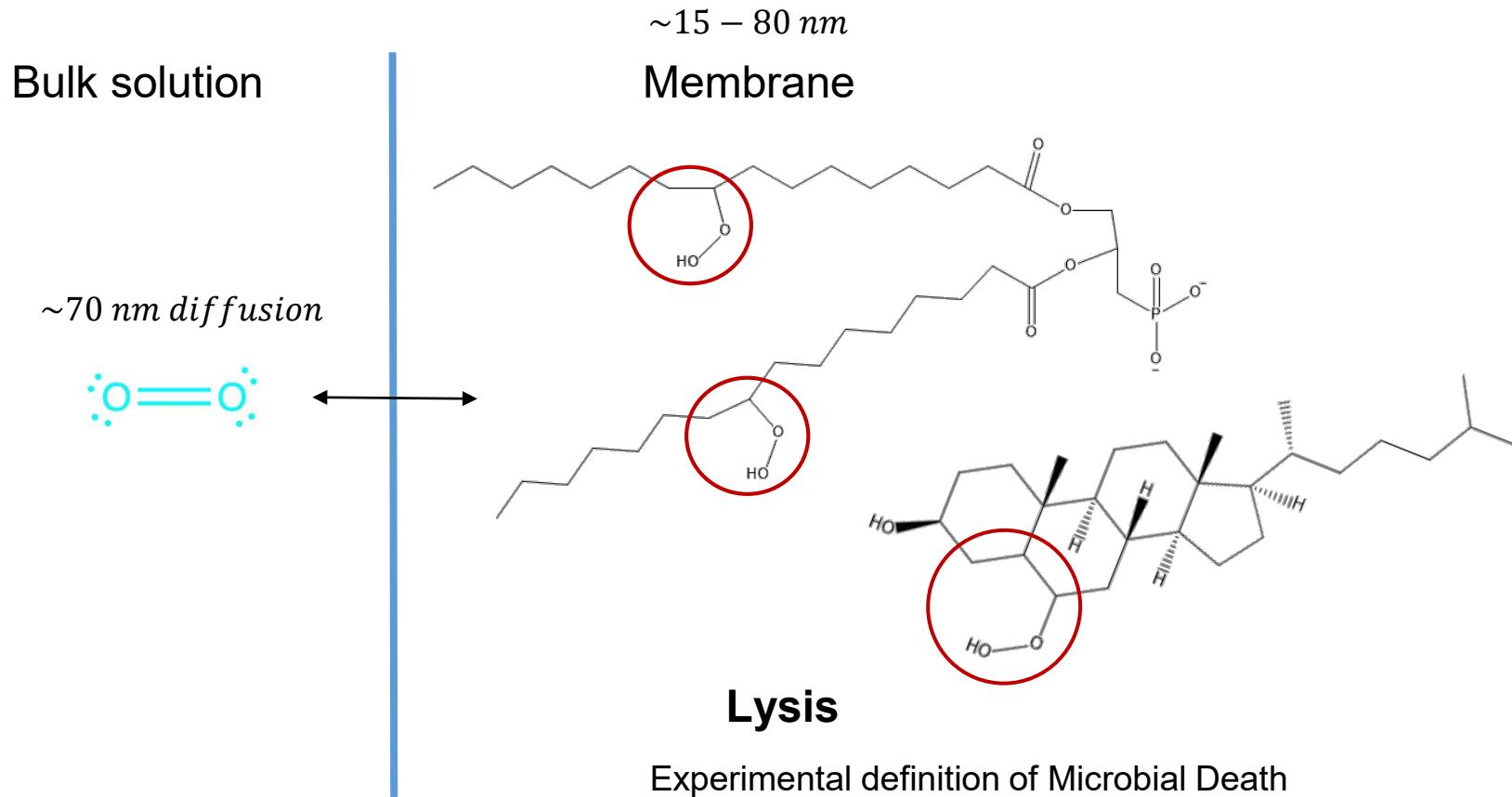
Human skin thickness $\approx 1.8 \text{ mm}$

Singlet oxygen layer ($\sim 75 \text{ nm}$)

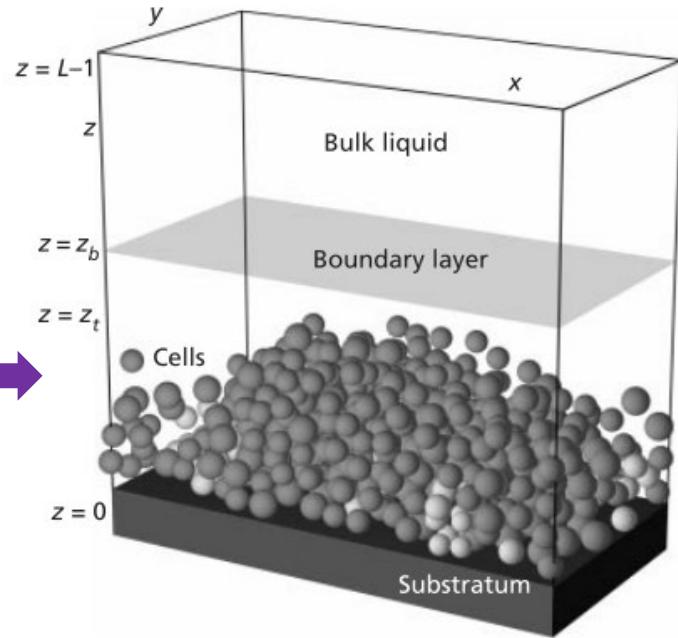
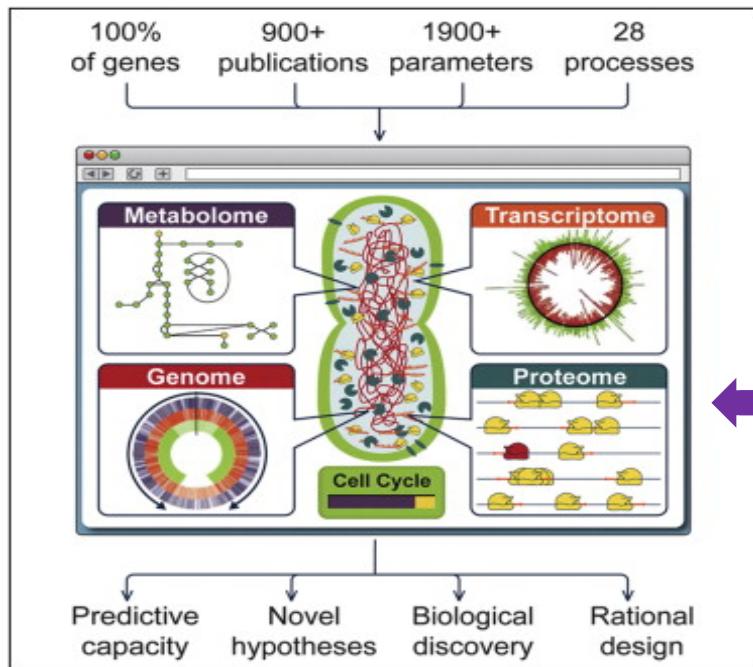
Polyamide RO filtration membrane ($\sim 150 \text{ nm}$)



Membrane inactivation



Computational inspiration



The Whole Cell Model (*Cell*, 2012)

- Bottom-up biochemical accuracy

Biofilm Models

- Top-down deterministic ODEs



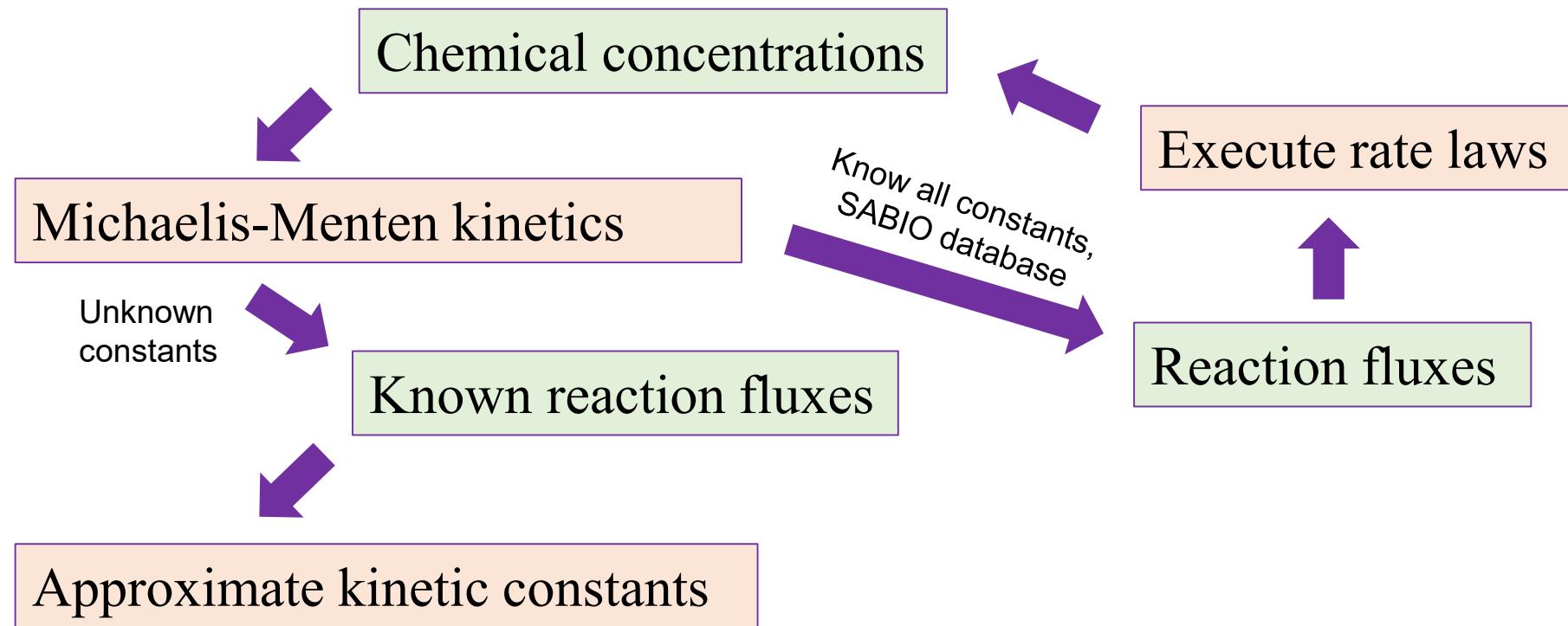
Model intentions

- 1) Predict results
- 2) Educate mechanisms



Green: quantity ; Orange: calculation

Kinetic workflow



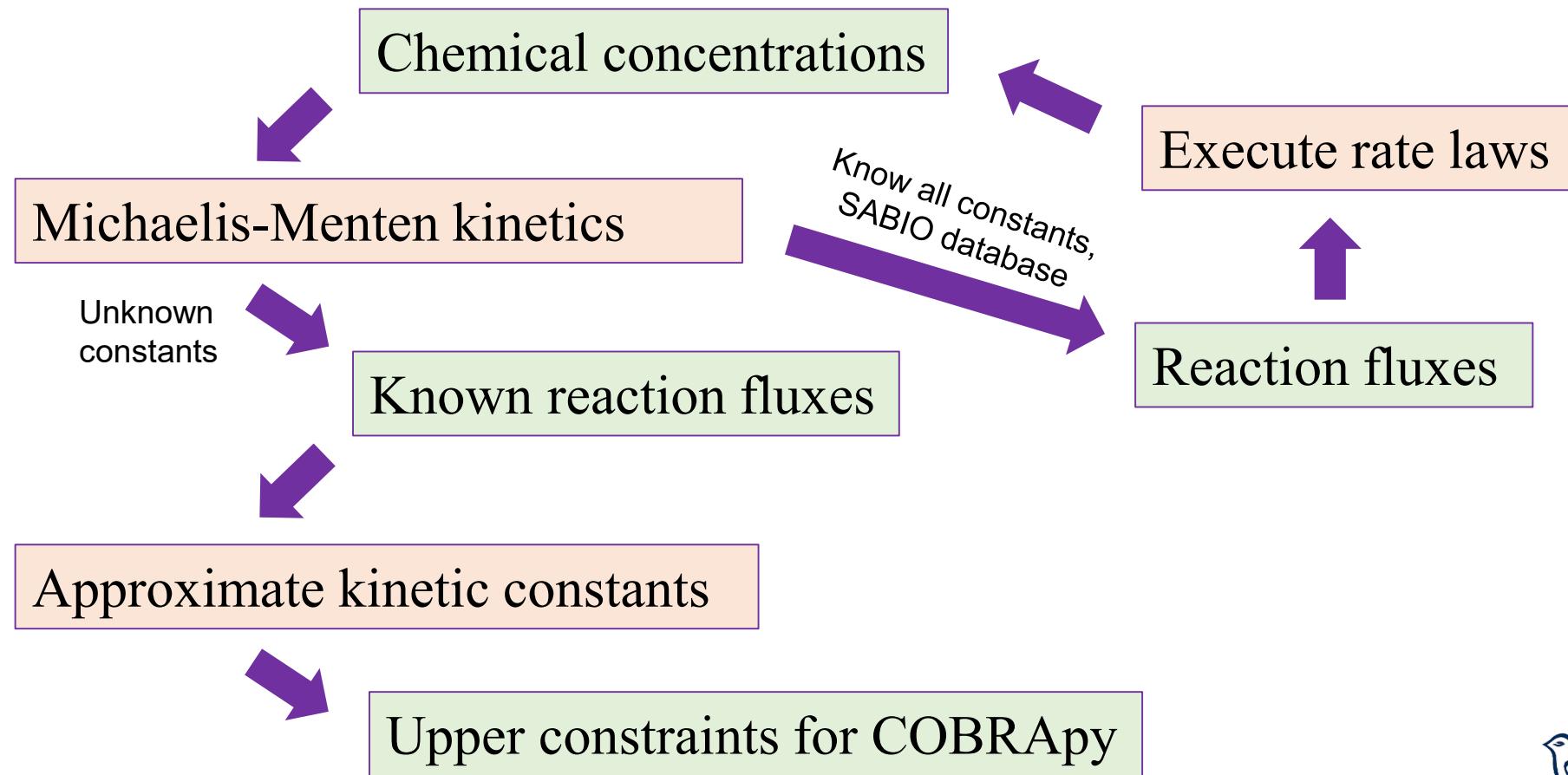
Kinetic approximations

- Quantitative structure-property relationships (QSPRs)
- Thermodynamic proxy
- Potentially negligible missing reactions



Green: quantity ; Orange: calculation

Kinetic workflow



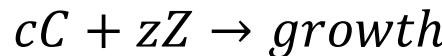
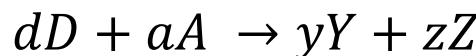
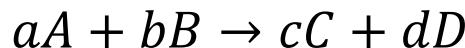
Chemical reaction dynamics “flux”

3) Flux balance analysis ([Cobrapy](#)) – linear programming toward a directive



$$\begin{bmatrix} -a & -a & 0 \\ -b & 0 & 0 \\ c & 0 & -c \\ d & -d & 0 \\ 0 & y & 0 \\ 0 & z & -z \\ 0 & 0 & \text{growth} \end{bmatrix} \cdot \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} = 0 \quad [\text{steady state}]$$

maximize growth = maximize v_3

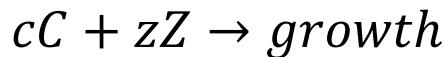
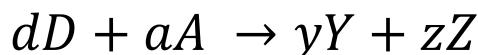
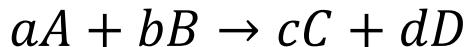


Chemical reaction dynamics “flux”

3) Flux balance analysis ([Cobrapy](#)) – linear programming toward a directive

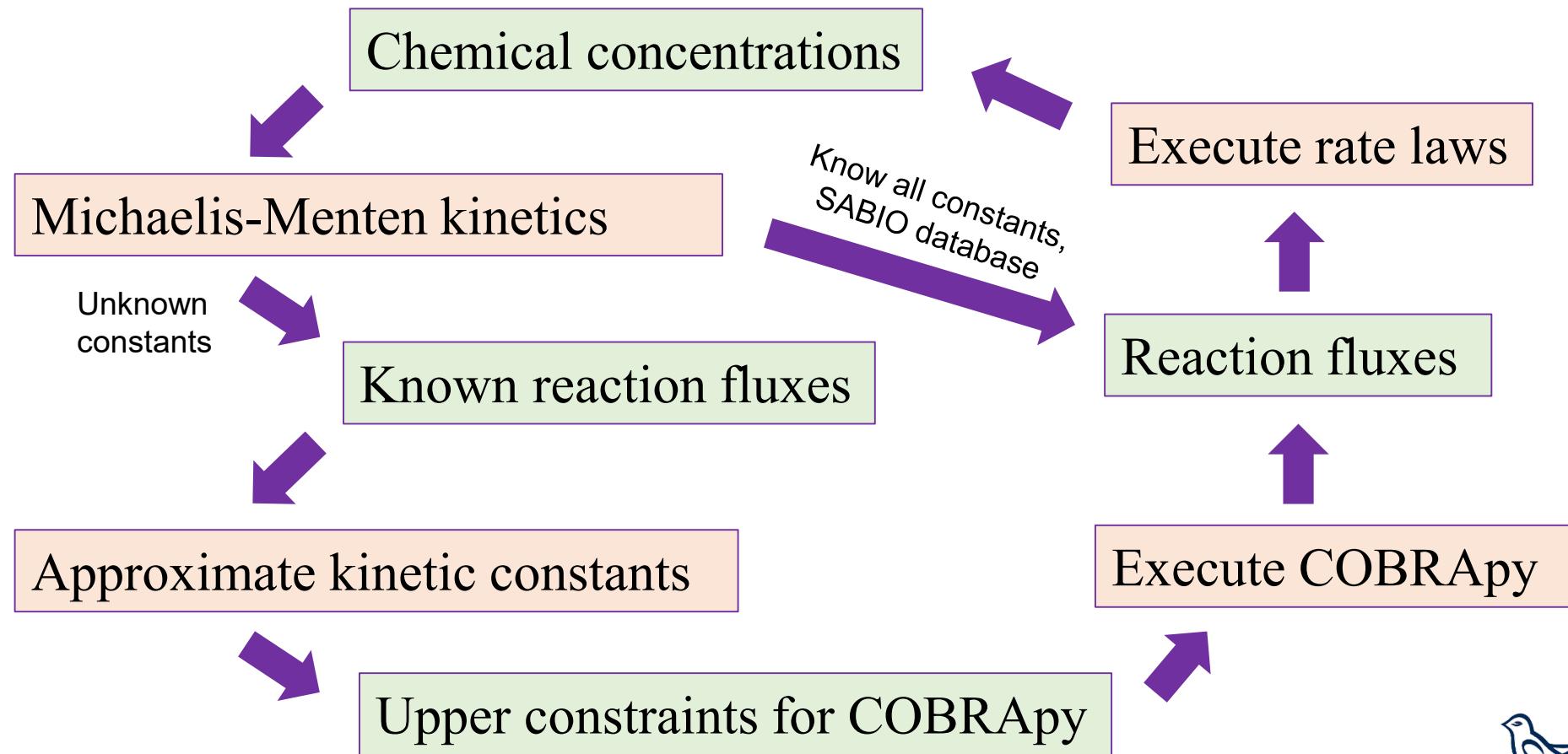
- Thermodynamic reaction limits are approximated
 - v of forward reactions $\subseteq [0,1000]$
 - v of reversible reactions $\subseteq [-1000,1000]$

$$bound_{lower} \leq v_a \dots \leq bound_{upper}$$



Green: quantity ; Orange: calculation

Kinetic workflow



Web scraping and database curation

1) Standardize biochemical databases

- *WholeCellKB.org*
- NIST *Thermodynamics of Enzyme-Catalyzed Reactions*
- *SABIO kinetic database*



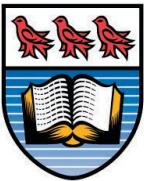
2) Partnership with the ModelSEED database.



New directions

- 1) Approximate the missing kinetic values
- 2) Execute a preliminary cellular model
- 3) Incorporate singlet oxygen and biofilm reactions
- 4) Visualize through results plots and a GUI





University of Victoria



Green Safe Water Lab



Mitacs
ACCELERATE

A scenic landscape of Lake Tahoe. The foreground shows clear, turquoise-blue water with several large, light-colored rocks at the bottom. A dense forest of tall, green pine trees lines the shore. In the background, there are more pine trees and a range of mountains under a bright blue sky with a few wispy white clouds.

¿Questions?

¿Suggestions?

Kinetic calculations

$$v_1 = \frac{v_{max1} * [A] * [B]}{(K_{M_a} + 1) * [A] + (K_{M_b} + 1) * [B]} ; v_{max1} = k_{cat1} * [Enzyme_1]_0$$

⋮

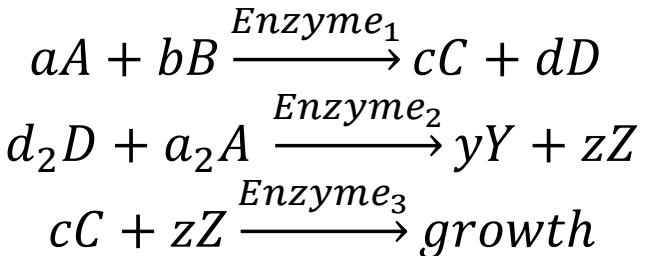
⋮

$$\frac{d[A]}{dt} = v_1 * -a + v_2 * -a_2$$

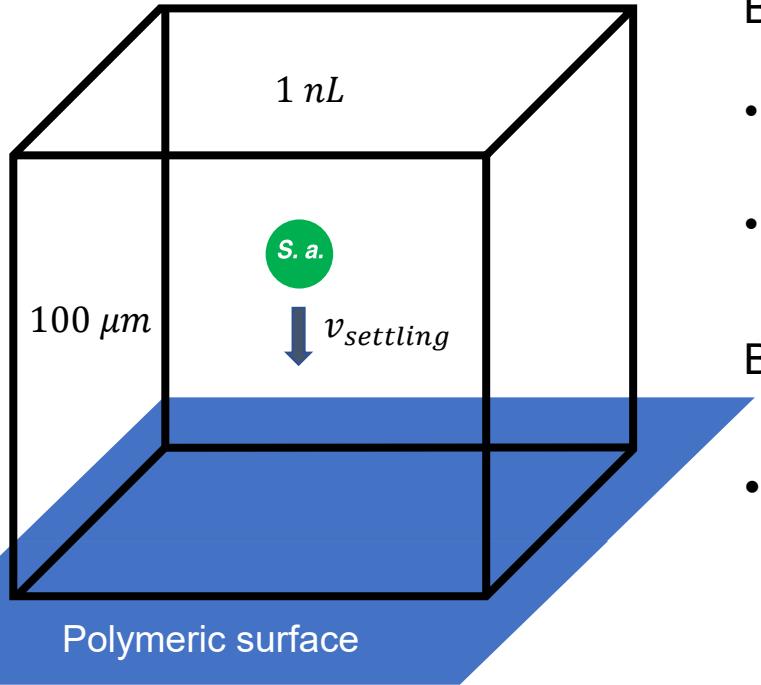
$$\frac{d[D]}{dt} = v_1 * d + v_2 * -d_2$$

⋮

⋮



Simulation space with settling



Bacterium $\approx 1 fL$

- 1 million grid volume cells
- $\frac{\text{bacterium}}{\text{simulation volume}} = \frac{1 fL}{1 nL} = \frac{1E-15 L}{1E-9 L} = (1E - 6) = ppm$

Bacterium = particle

- Stoke's law of terminal velocity

- $v_{settling} = \frac{g * (\rho_{bacterium} - \rho_{water}) * d_{bacterium}^2}{18 * \mu}$
 $\approx 0.25 \frac{\mu m}{s}$

- $Re = \frac{\rho_{water} * d_{bacterium} * v_{settling}}{\mu} \approx 4E - 9 \ll 2$

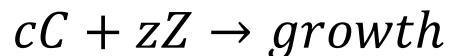
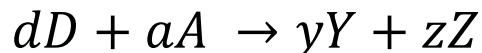
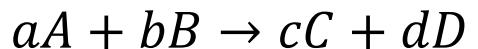


Chemical reaction dynamics “flux”

3a) Dynamic FBA ([dfba](#)) – Time variability and dependence

- $[A], [B], [C], [D], [Y], [Z]$ are variable over simulation time

```
Set the initial biomass concentration
Set the initial conditions of the environment
From the starting time to the final time
    Based on the current biomass concentration and environmental conditions,
        set the upper and lower bounds of the exchange reactions
    Solve for the maximum growth rate and optimal fluxes
    Update the biomass concentration based on the predicted growth rate
    Update the environmental conditions based on the predicted exchange fluxes
```



User-defined parameters

1) Extracellular conditions

- LB broth, as casein and yeast extract ; temperature ; NTUs, et cetera

Download COBRA model from the BiGG Database:

1 to 108 (108)

SBML [?](#) : RECON1.xml (.xml.gz, compressed)
JSON [?](#) : RECON1.json (.json.gz, compressed)
MAT [?](#) : RECON1.mat (.mat.gz, compressed)

2) Bacterium species

- *Mycoplasma genitalium* => cell cycle, cell mass\volume, metabolic proportions

3) Inactivation method

- Photodynamic inactivation (PDI) with singlet oxygen
- Organic antimicrobial agents



Approximated parameters for *M. genitalium*

- 1) $\text{starvation proportion} = \left(\frac{\text{mass}_t}{\text{mass}_{t=0}} \right) = \frac{1}{3}$ *Quite uncertain
 - The fraction of initial mass below which the bacterium dies
- 2) ** $\text{translation rate} = 4 \left(\frac{\text{codons}}{\text{second}} \right)$
 - The rate at which codons are transcribed into amino acids for protein synthesis
- 3) ** $\text{enzyme halflife} = 5000 \text{ (seconds)}$
 - The rate of enzyme degradation into amino acids for every enzyme
- 4) $\text{bacterial, electrical cell potential} = 0.2 \text{ (volts)} ^*$
 - Electric cell potential of the bacterium, which is the aggregation of all biochemical reactions
- 5) $T_{opt} = 310 \text{ }^{\circ}\text{K}$
 - The optimum incubation temperature, which is the aggregate thermodynamics of the bacterium
- 6) ** $\text{reaction completeness} = \frac{\text{reactions executed}}{\text{possible reactions}} = 0.9$
 - The proportion of necessary reactions to achieve $\left(\frac{Q}{K_{eq}} \right)_{opt}$ that are executed in a timestep



Approximated parameters for *M. genitalium*

7) *Planktonic* = 1 ; *Detached* = 0.8 ; *Biofilm* = 0.5 ; *Persister* = 0.00001

- The relative metabolic reaction rate, which is informed through an interview of a Eukaryotic biologist

8) ** *vital energetic proportion* = $\frac{(\text{energetic proportion})_t}{(\text{energetic proportion})_{t=0}} = \frac{1}{2}$

- The proportion of energetic chemicals relative to below which the bacterium dies



Assumptions and limitations

- 1) Homogeneous bulk and cytoplasm
- 2) Mass balance applies everywhere
- 3) ** Need-based absorption
 - $need = \left(\frac{Q}{K}\right)_{optimum} - \left(\frac{Q}{K}\right)_{current}$
 - $\left(\frac{Q}{K}\right)_{optimum}$ is estimated from the incubation °K
- 4) ** Boltzmann distribution to substrates
- 5) ** Transcription and/or enzymes are negligible
- 6) Singlet oxygen oxidizes only unsaturated lipids
- 7) Constant cell cycle times
 - Only three phases: Interphase, S, and mitosis
- 8) Replication resets the bacterium
- 9) $Q_{bacterium} = constant$
- 10) The acquired datasets are accurate



System dynamics

Whole Cell Biofilm Model

Planktonic-phase bacteria

Bacterial growth & replication

Bacterium

Metabolism

Nutrients

Anti-biotics

Biofilm Growth Models

Conditional layer

Quorum sensing

Dispersion

Diffusion

Detachment

Channels/pores

Shear forces

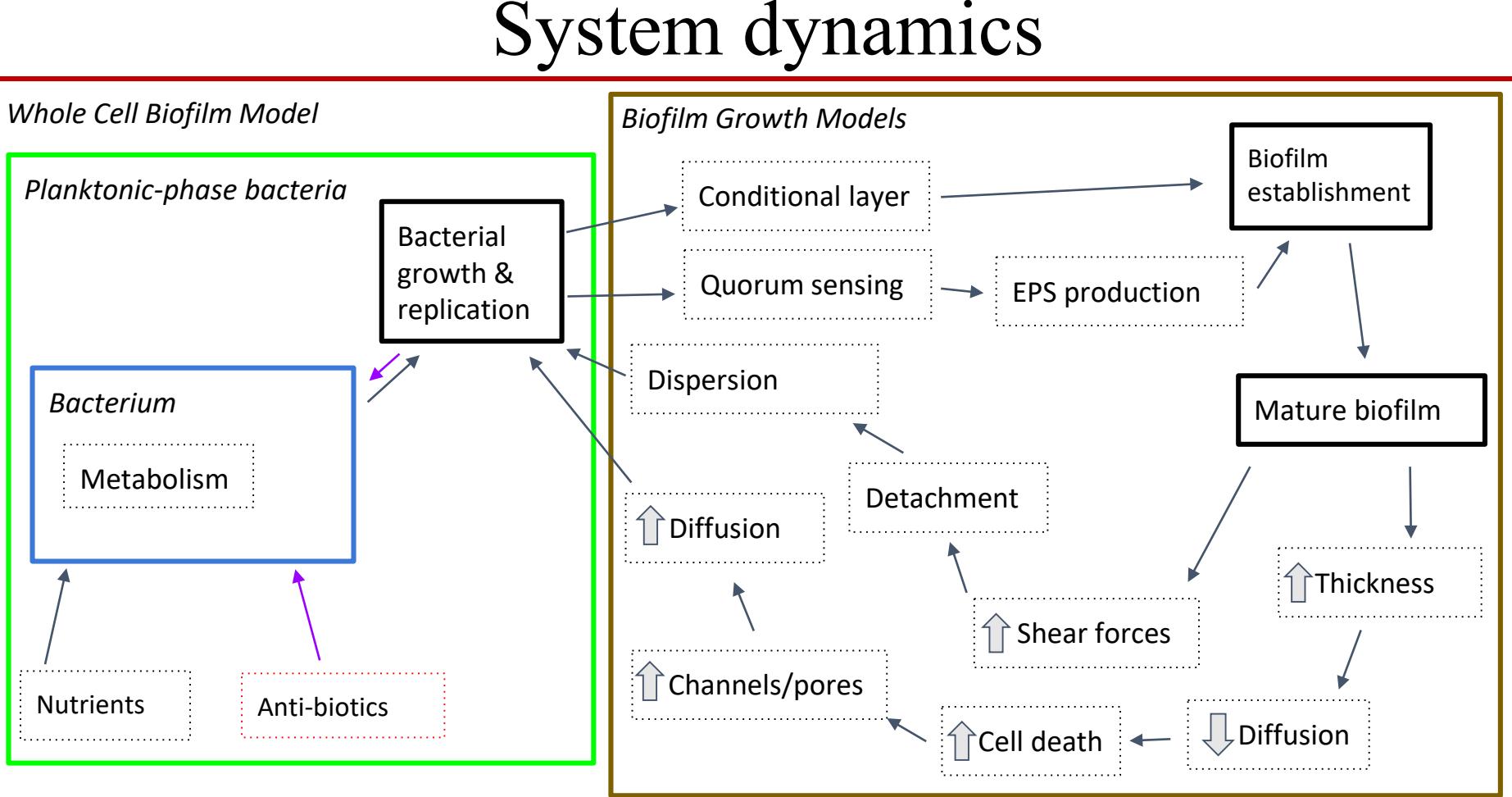
Cell death

Biofilm establishment

Mature biofilm

Thickness

Diffusion



** Membrane flux = mass balance

$$mass_{bacterium,t} = mass_{t=t-1} + mass_{net,t}$$

$$mass_{net,t} = mass_{absorbed,t} - mass_{ejected,t}$$

$$mass_{absorbed,t} = \sum_{i=1}^a n_{i,absorbed} * MW_i$$

a = # absorbed molecules

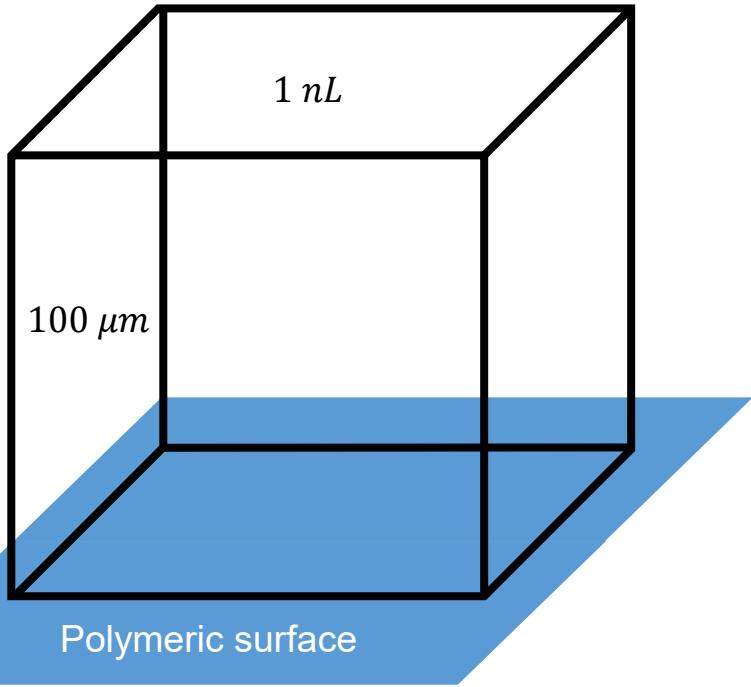
$$mass_{ejected,t} = \sum_{i=1}^z n_{i,ejected} * MW_i$$

z = # ejected molecules

$$V_{bacterium,t} = m_{bacterium,t} * \left(\frac{V_{bacterium,0} \approx 1 fL}{m_{bacterium,0} \approx 1 pg} = constant \right)$$



System boundaries



Bacterium $\approx 1 fL$

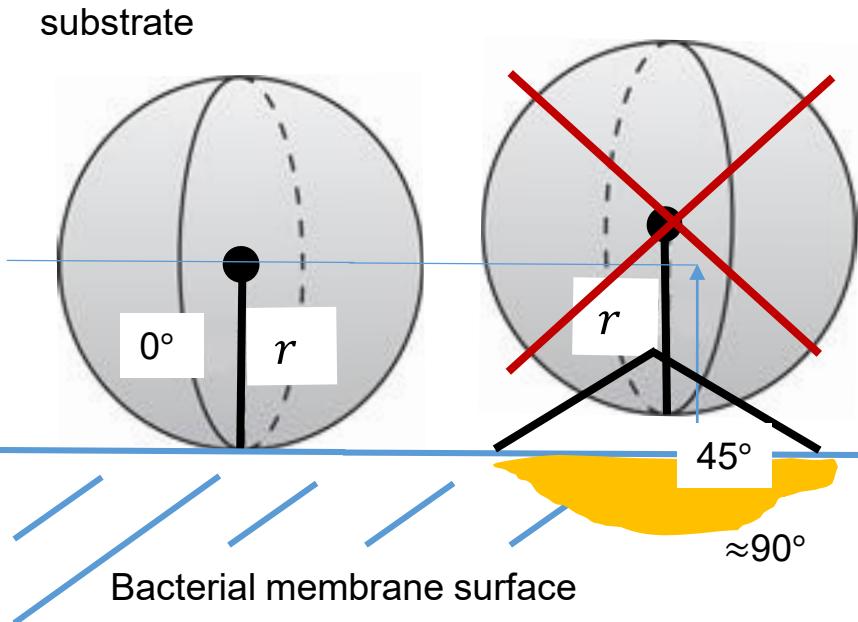
- 1 million grid volume cells
- $$\frac{\text{bacterium}}{\text{simulation volume}} = \frac{1 fL}{1 nL} = \frac{1E-15 L}{1E-9 L} = (1E - 6) = ppm$$

$$\bullet \quad (\#viable\ bacterium)_{t=0} \approx 2E6 \left(\frac{CFU}{mL} \right) = 2E9 \left(\frac{CFU}{L} \right) = 2 \frac{CFU}{simulation} = 2 CFU ppm$$

$$\bullet \quad (\#viable\ bacterium)_{t=final} \approx 2E9 \left(\frac{CFU}{mL} \right) = 2E3 \frac{CFU}{simulation}$$



** Proportion of interactions from bulk



\overrightarrow{V}_{rms} = rms velocity of substrates

$$r = \overrightarrow{V}_{rms} * \Delta t$$

Surface area = probability

r = maximal distance for a substrate interaction in Δt
 ≈ 0 probability of membrane interaction

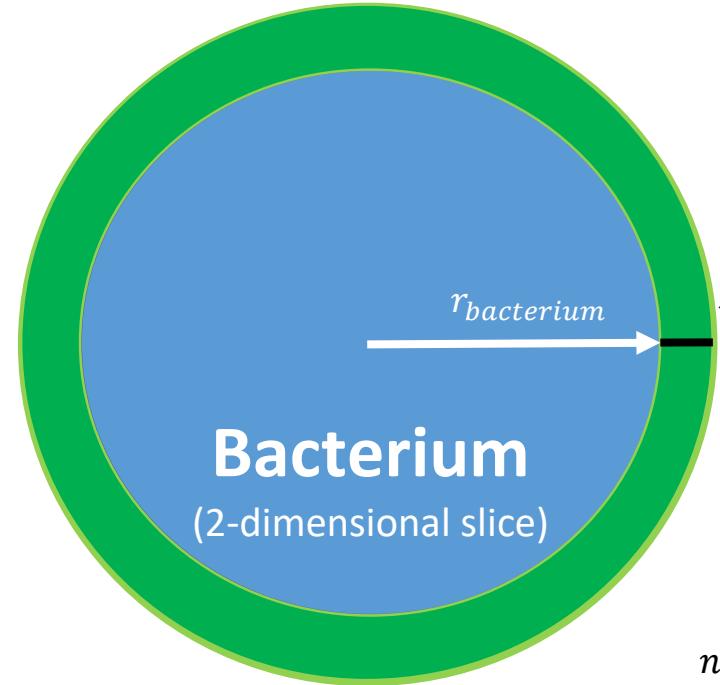
$\approx \frac{1}{2}$ probability of membrane interaction

$\approx 45^\circ$ average angle between 0° and $\approx 90^\circ$

$$\text{average probability of interaction} = \bar{P} = \frac{\text{surface area (intercepted spherical cap)}}{\text{surface area (sphere)}} = 14.6\%$$



** Chemical adsorption



C_i = concentration of bulk substrate i
 x = # reactions with i
 β_i = barrier inhibition
 y = # substrates i

$$m_{\text{absorbed}} = \sum_{i=1}^y n_i * MW_i$$

$$n_{i,\text{absorbed}} = \begin{cases} n_{i,\text{interaction}} * \frac{\left(\frac{Q}{K_{eq}}\right)_{\text{optimum}} - \Pi_{i=1}^x \left(\frac{(Q_i)_{t,C_i}}{(K_i)_{eq,C_i}}\right)}{\left(\frac{Q}{K_{eq}}\right)_{\text{optimum}}} * \beta_i & \left(\frac{Q}{K_{eq}}\right)_{\text{optimum}} - \Pi_{i=1}^x \left(\frac{(Q_i)_{t,C_i}}{(K_i)_{eq,C_i}}\right) > \left(\frac{Q}{K_{eq}}\right)_{\text{optimum}} \\ 0, & \Pi_{i=1}^x \left(\frac{(Q_i)_{t,C_i}}{(K_i)_{eq,C_i}}\right) > \left(\frac{Q}{K_{eq}}\right)_{\text{optimum}} \end{cases}$$

$$V_{\text{shell}} = \left(\frac{4\pi}{3}\right) * (r_{\text{outer}}^3 - r_{\text{bacterium}}^3),$$

with $r_{\text{outer}} = r_{\text{bacterium}} + r_{\text{interaction}}$

$$r_{\text{interaction}} = \overrightarrow{V}_{rms} * \Delta t$$

$$n_{i,\text{interaction}} = [C]_i * V_{\text{shell}} * P$$



** Optimal thermodynamics

$$\left(\frac{Q}{K_{eq}}\right)_{optimal} = e^{\frac{-n*F*E}{R*T_{opt}}}$$

ΔG = Gibbs free energy

ΔG^0 = Gibbs @ standard conditions

T_{opt} = optimal bacterial incubation temperature

R = gas constant

Q = reaction quotient = $\frac{[product]^n}{[reactant]^m}$

$$n = <\frac{e^-}{mol}>$$

F = Faraday's constant

E = electrical potential of the bacterium

K_{eq} = equilibrium constant = $\frac{[product]^n}{[reactant]^m}$

$$1) \Delta G = \Delta G^0 + R * T_{opt} * \ln(Q)$$

$$2) \Delta G = -n * F * E$$

$$3) \Delta G^0 = -R * T_{opt} * \ln(K_{eq})$$

$$-n * F * E = R * T_{opt} * (\ln Q - \ln K_{eq})$$

$$\left(\frac{-n * F * E}{R * T_{opt}} \right) = \ln \left(\frac{Q}{K_{eq}} \right)$$



New directions

1) Thoroughly organize the reaction database

- Categorizing reactions as inter-\intra-compartmental

2) Expand biochemical accuracies

- Introduce quorum sensing reactions

3) Expand the bacterium model into a biofilm model

- Incorporate new functionalities like metabolic states

4) Compare with conventional methods

- Flux balance analysis via cobrapy module

5) Introduce a visual depiction

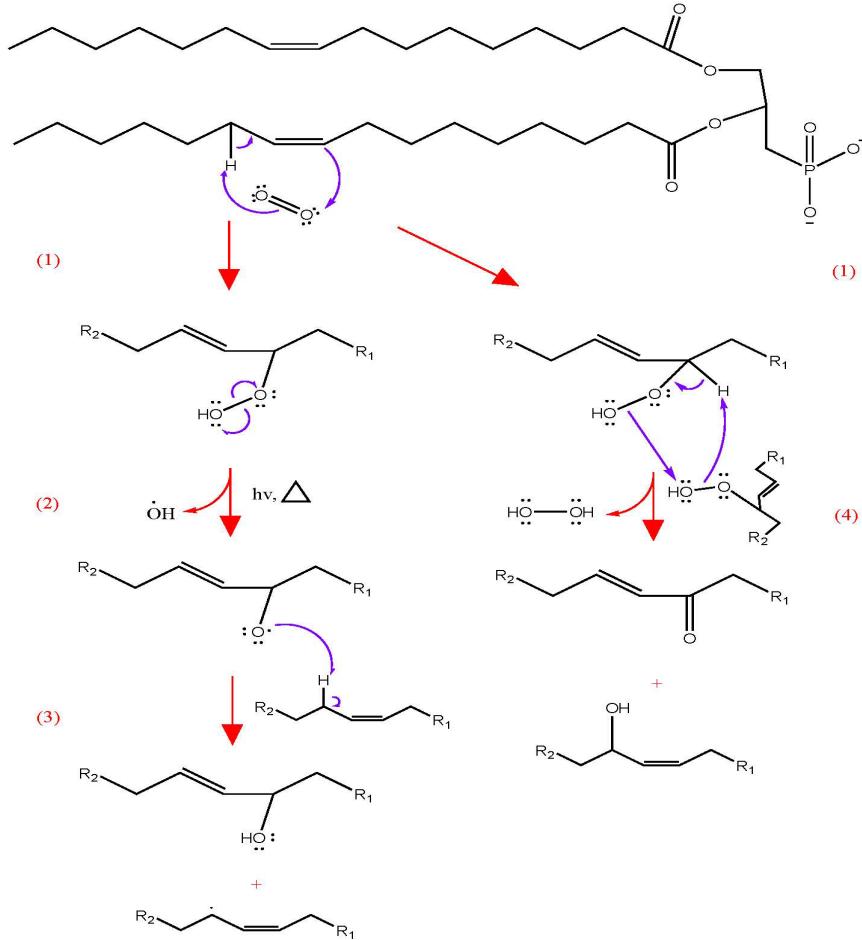
- Matplotlib plots and tkinter GUI

6) Implement antibiotic reactions

- Parameterize ${}^1\text{O}_2$ reactions



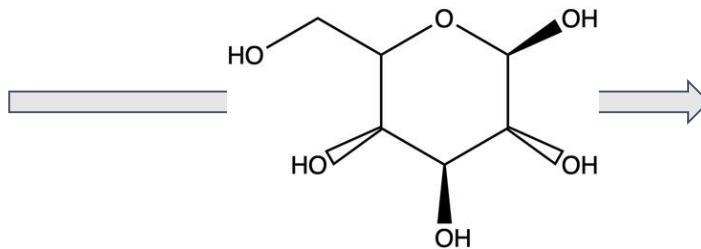
Membrane oxidation mechanisms



Suspension phase model

Solution elements

Substrate



Parameters/equations

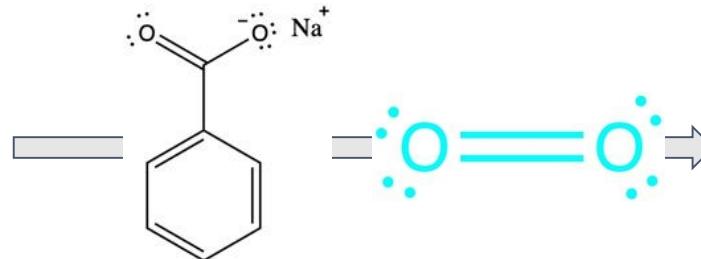
Diffusion - Reaction-diffusion Equations
Fick's law or Cahn-Hilliard equations

Water



Diffusion - Reaction-diffusion Equations
Fick's law or Cahn-Hilliard equations

Anti-foulants



Disrupt Biochemistry -
Quorum sensing
Bacteriostatic
Bactericidal
Diffusion - Reaction-diffusion Equations
Fick's law
Cahn-Hilliard equations



Modeled biofilm and membrane

Biofilm qualities

Shape



Parameters/equations

Shear forces - Digital Biofilm model (2016)
Thickness - Biofilm Growth Model (2000 MSU)

Channeled



Porous - Cellular automata algorithm

Membrane



Porous - Biofilm Growth Model (2000 MSU)



Modeled bacteria



Bacterial elements

Parameters/equations

Motility



Staphylococcus aureus is not motile

Density limit



Quorum sensing - Frederick et al. 2016
Daughter cell dispersion - Individual-based algorithm

Bacterial growth



Substrate - Monod kinetics
- Michaelis-Menten kinetics
Anti-foulant - **Novel**



Biofouling

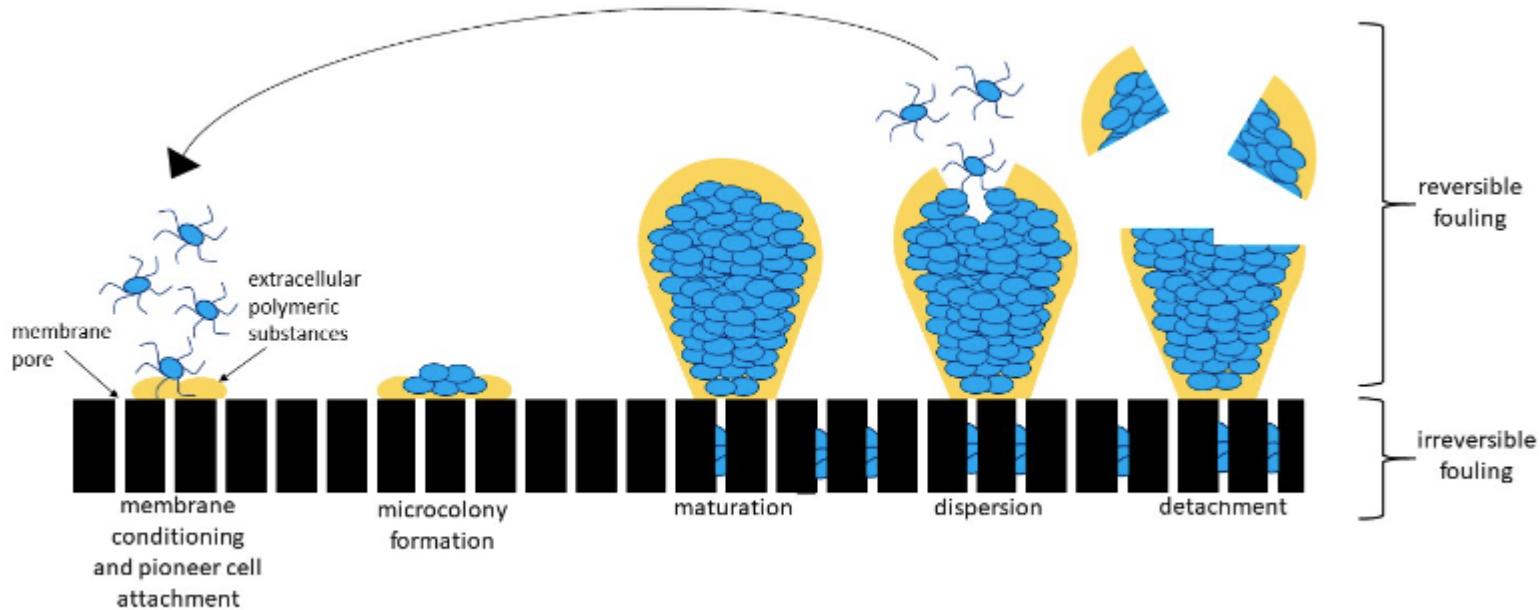
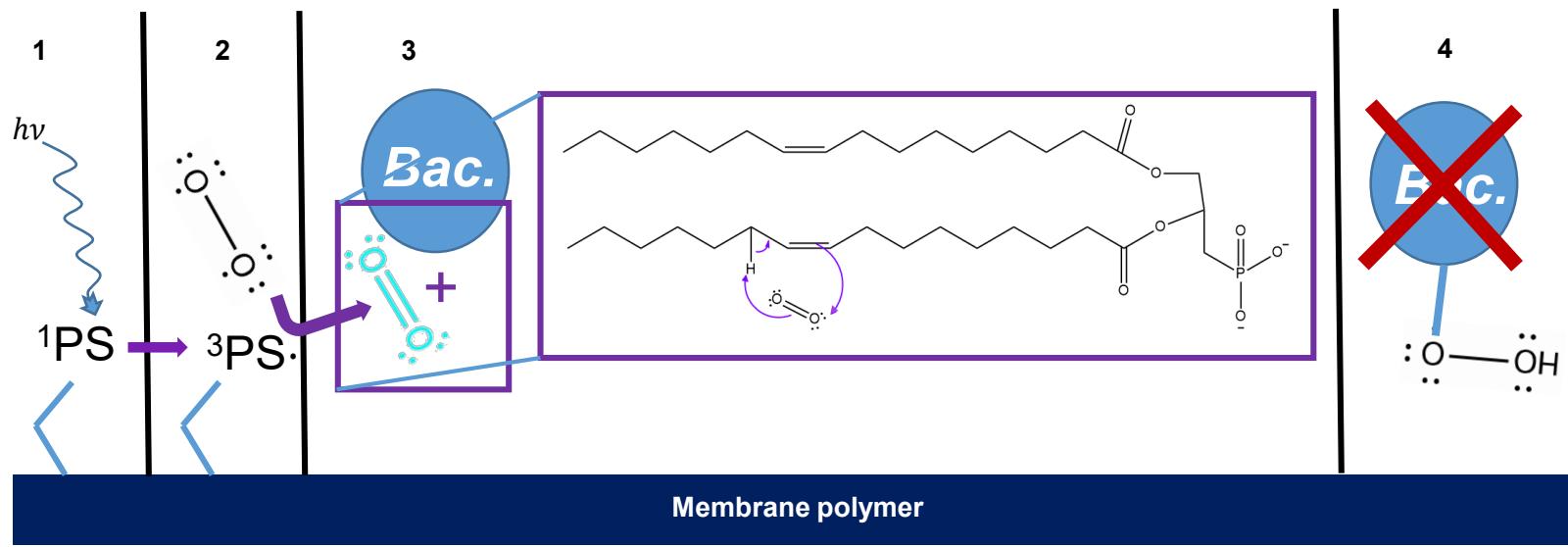


Figure 1. Biofilm growth on a semipermeable membrane (Image Credit: Anna Curtin)



Fundamental questions



Algorithm/Model	Assumptions/limitations	Model contribution
<i>Rittmann model & Biofilm Accumulation model (BAM)</i>	1) The biofilm is composed of dead+active bacteria and water 2) Constant Biofilm growth, [Substrate], and bulk volume	Foundation
<i>Biofilm Growth model (BGM)</i>	1) The biofilm is composed of dead+active bacteria and water 2) Constant Biofilm growth and [Substrate]	Dynamic bulk volume
<i>Digital Biofilm Model (DBM)</i>	1) Biofilms are two-phases: rigid bacteria and malleable EPS 2) Proteins were only modeled in the EPS	Accurate Biofilm composition
<i>Individual-based algorithm</i>	1) Computational demands 2) Bacteria are inelastic spheres 3) Porosity is predestined by net vector daughter cell dispersal	Natural evolution of population growth
<i>Cellular automaton algorithm</i>	1) Heterogeneous bacteria and biofilm 2) Unrealistic quantization of parameters 3) Parameters values can be subjective	Mature biofilm channelling

Whole Cell Biofilm Model



Algorithm/Model	Assumptions/limitations	Model contribution
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Whole Cell Biofilm Model



Cellular Automaton algorithm

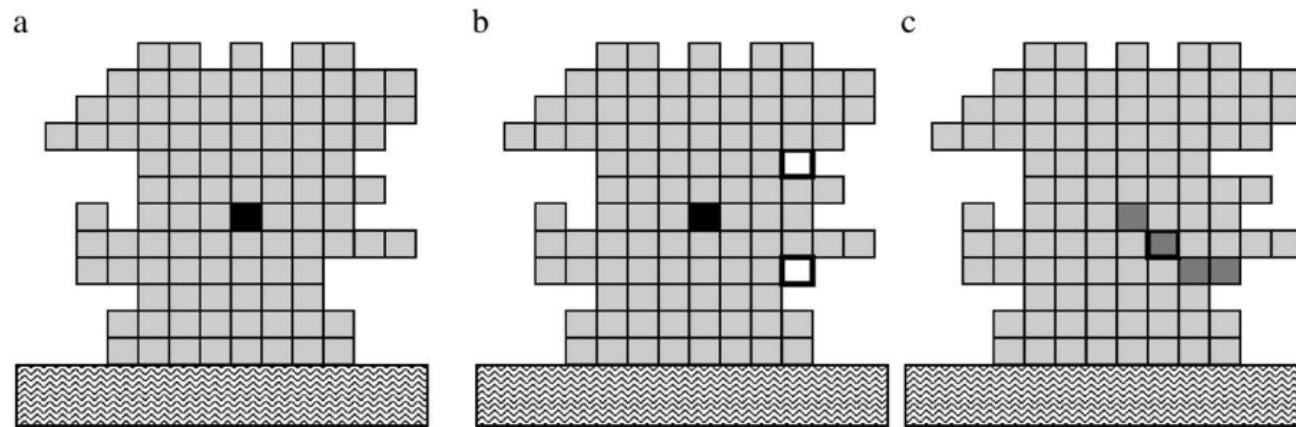
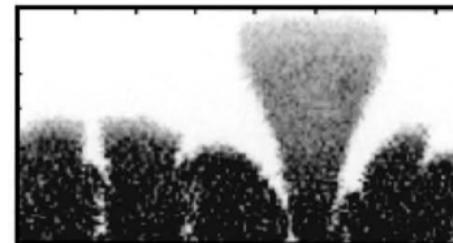


Fig. 2. The CA algorithm used in the UMCCA model. (a) Overflowing compartment, (b) The nearest two equidistant compartments to the overflowing compartment, (c) The algorithm randomly picks one of the two and places the new biomass in a neighboring compartment (bold compartment), while it shoves existing biomass along the path of least resistance.

Cellular Automaton algorithm

- Lattice Cartesian grid
- Stochastic selection of the closest unoccupied cells



Individual-based algorithm

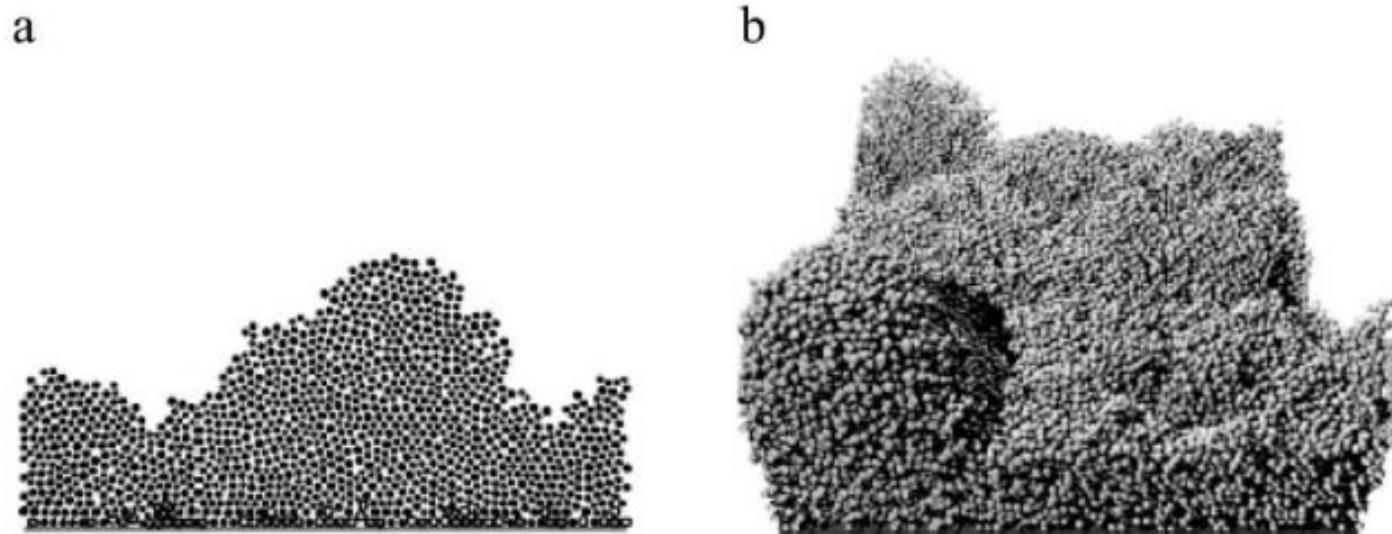


Fig. 3. Sample output of the IbM by Picioreanu et al. [29] in (a) 2-D and (b) 3-D. (c) sample 2-D output

Individual-based algorithm

- Dispersal according to the net vector from cellular overlap



Bioassays

