

Background

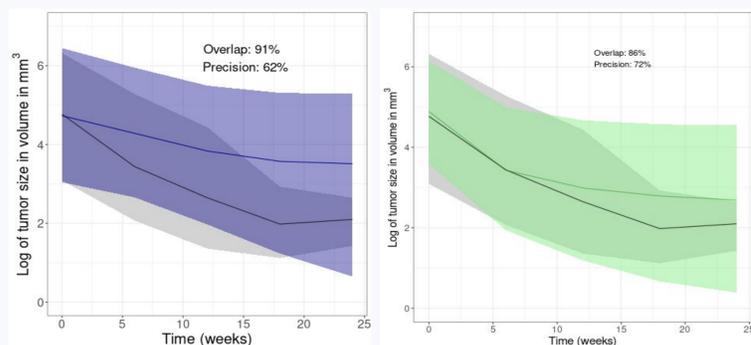
Treatment for lung adenocarcinoma (LUAD) patients is complex: mechanistic modeling based on knowledge could help clinicians testing biological assumptions, and comparing and combining treatments, if tumor evolution in size and genetic heterogeneity can be represented. We therefore present a knowledge-based mathematical mechanistic model that could complement clinical decision.

Model Benchmark

Protocol

We ran an *in silico* trial following 4000 EGFR+ LUAD patients, treated with gefitinib 250mg once daily, and monitored their tumor size, as reported in Nagase *et al.*'s [4] model.

- Overlap: model's ability to reproduce real life variability
 $O = (\text{Obs. interval} \cap \text{Pred. interval}) / \text{Obs. interval}$
- Precision: model's ability to provide results with reasonable variability
 $P = (\text{Obs. interval} \cap \text{Pred. interval}) / \text{Pred. interval}$



ISELA and Bayesian models against raw data are plotted. For each distribution (raw data ; ISELA model ; Bayesian model), the median and 25-75% interval is provided and used for computation of overlap and precision. Grey: raw data reported by Nagase *et al.*; Blue: ISELA model; Green: Bayesian model.

Next steps?

- Challenge the model in reproducing clinical data not used for its construction nor calibration
- Extend the ISELA model to increase its context of use beyond EGFR+ LUAD patients

Project Formulation

LUAD: most common lung cancer subtype [1]. About 1/3 of LUAD harbor a tumor driver mutation in Epidermal Growth Factor Receptor (EGFR+) [2]. SoC for advanced EGFR+ LUAD is EGFR tyrosine kinase inhibitor, e.g. gefitinib [3].

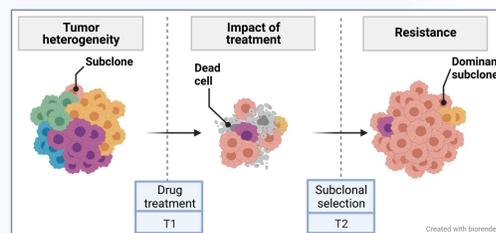


Illustration of the phenomena that drive tumor growth. Cell colors represent tumor subclones: different genotype but similar phenotype

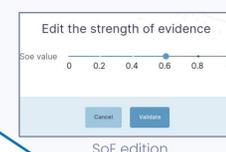
Aim of the study: produce a mechanistic model of lung cancer to predict tumor size evolution over time in EGFR+LUAD patients, and compare it to an already published model [4].

Results

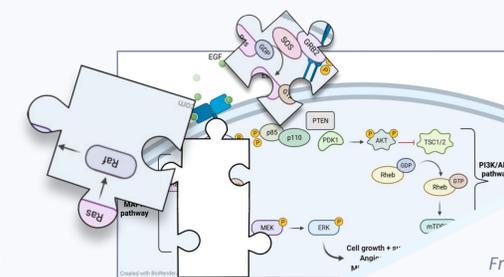
Both models efficiently reproduce experimental data, indicating that ISELA model is able to reproduce the evolution of tumor size of a database not used to build or calibrate it.

Systemic Knowledge Review

- 300+ scientific papers
- Identify the relevant biological entities and their relationship
- Define a Strength of evidence (SoE) scoring for every piece of knowledge
- Manual collection, selection, evaluation of pieces of knowledge



Realized on the **jinkō platform**, a state-of-the-art clinical trial simulation cloud-based SaaS solution.



We have assembled pieces of knowledge

Computational Model Development

From knowledge to mathematical equations, embedded in computer code

Mathematical modeling

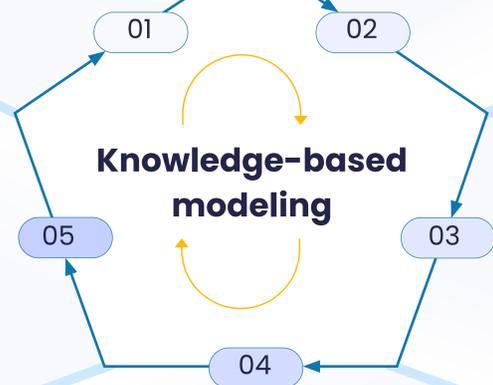
5 biological models implemented in parallel & integrated with connecting variables. The *in silico* EGFR+ Lung Adenocarcinoma (ISELA) model is composed of up to 83 ordinary differential equations (ODEs), 43 variables, and 170 parameters reflecting intra-tumor heterogeneity.

Focus on the clonal heterogeneity

- Tumor subclone:** Group of tumor cells that might harbor genotype differences but that share the same phenotype. In the model, we consider:
- Coexistence of 2 to 15 subclones [5]
 - Preserved EGFR mutation across subclones
 - Subclones with similar behavior in response to biological signals
 - Subclones with distinct properties derived from sub-clonal mutations
 - Inter-subclonal variability due to mutations, or epigenetic alterations
 - Some yield a new tumor subclone

Relevant model endpoint

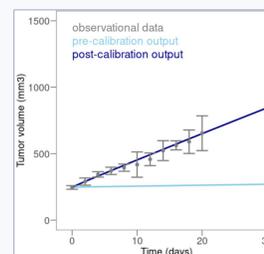
Tumor radius: Radius in centimeter of the primary tumor, under the assumption that the tumor is a spheroid.



Calibration & Virtual population

Calibration process

1. Experimental data gathering
 - From 26 published experiments: spheroids, xenografts, clinical results, with relevant mutational profiles
2. Ensuring model adequation to experimental data
3. Step-by-step calibration protocol definition, composed of blocks with increasingly broad objectives
4. Fitness functions *f* creation representative of the experimental data
 - Iterative optimization until *f*-minimization is deemed satisfying

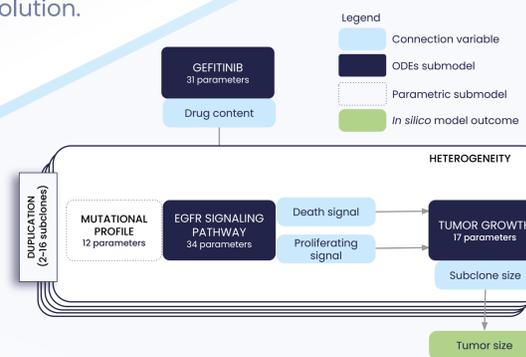


Example of simulation outputs before and after calibration (in this case tumor radius over time)

Virtual population for benchmark purposes

Virtual patients creation: vectors of values of parameters to respect the characteristics of the population used to build Nagase *et al.*'s model (2020) [4].

Characteristics	Nagase <i>et al.</i> 's model	ISELA model
Age, years (range)	59 (27 - 82)	58.8 (27.5 - 81.9)
Gender, % (n)		
- Women	81.8	82.8 (3312)
- Men	18.2	17.2 (688)
Asian, % (n)	99.7	100 (4000)
Common EGFR mutations % (n)	96.2	100 (400)



Graphical representation of the ISELA model. It includes 6 interconnected submodels of distinct granularity levels: from molecular reactions to tissue level:

- mutational profile,
- EGFR signaling pathways
- tumor growth
- gefitinib pharmacometrics
- tumor heterogeneity and duplication

References

- [1] Siegel *et al.*, CA CANCER, 2020
- [2] Vallee *et al.*, Int J Oncol, 2013
- [3] https://www.accessdata.fda.gov/drugsatfda_docs/label/2018/206995s003lbl.pdf
- [4] Nagase *et al.*, CPT Pharmacometrics Syst Pharmacol, 2020
- [5] Hanjani *et al.*, N Engl J Med, 2017

Acknowledgments

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Conclusion

The results obtained from the benchmark analysis ensure that the ISELA mechanistic model can reproduce tumor growth variability in advanced EGFR+ LUAD patients. The ISELA model is currently being expanded to clinical endpoints evaluation such as time to progression, in order to provide a tool for cost-effective research and development, to eventually overcome:

- reproducibility crisis, as we would be able to simulate an unlimited number of patients
- ethical and practical difficulties for control arms: modeling digital twins for each patient to effectively creates a synthetic control arm.

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