

Epidemic Engine Cloud-based API: Bridging Epidemic Simulators (GLEAM) with AI Algorithms



(etcao@ucsd.edu)

Alaa Fadhlallah (afadhlallah@ucsd.edu)

Manav Jairam (mjairam@ucsd.edu)

Anirudh Indraganti (aindraga@ucsd.edu)

Liam Manatt (Imannat@ucsd.edu)

Kyla (Dawon) Park (dap006@ucsd.edu)

Mentor: Rose Yu (roseyu@ucsd.edu)

Mentor: Yian Ma (yianma.ucsd@gmail.com)

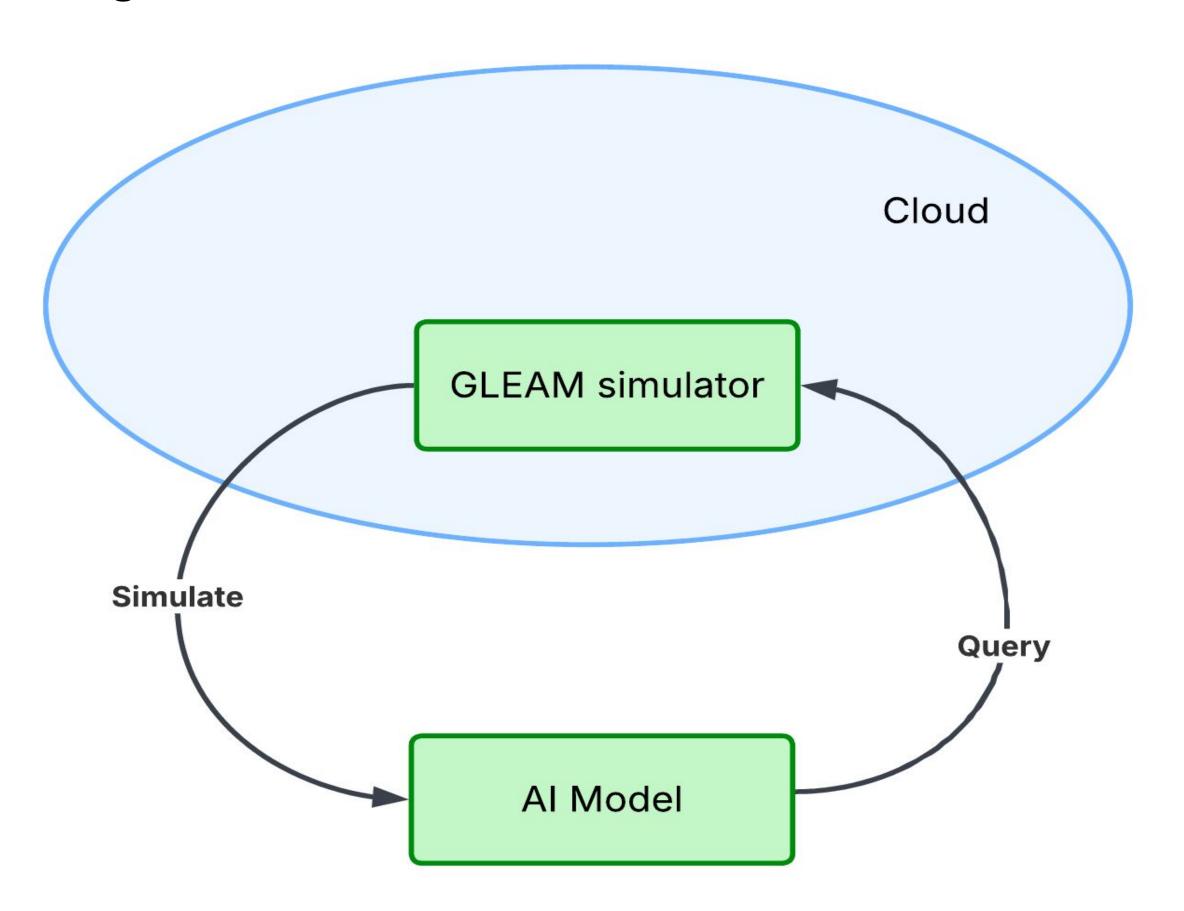
Background

Epidemic simulators model disease spread, but are computationally intensive.

Al surrogate models significantly reduce simulation time from weeks to days. However, current AI models rely on pre-simulated data, which is costly to store and lacks adaptability to evolving disease dynamics.

Research Goal

To develop an Epidemic Engine Cloud-based API service for real-time epidemic simulation framework that integrates traditional epidemic models with Al surrogate model acceleration.



Methodology: Model

GLEAM Model Integration:

Spatial, stochastic epidemic simulation uses real-world population and mobility data to generate high-fidelity data for training surrogate model.

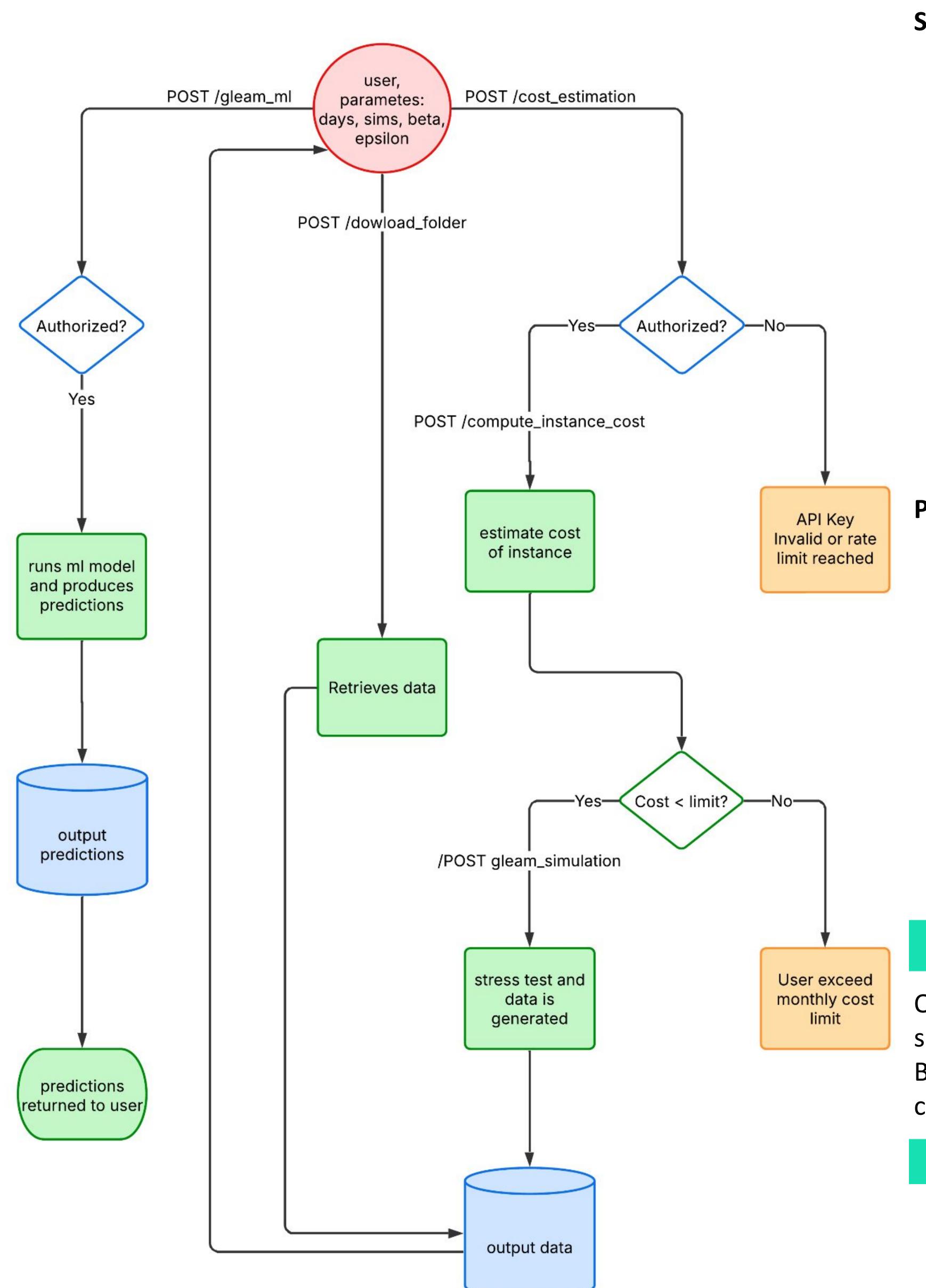
Bayesian Active Learning (BAL):

Uses latent information gain (LIG) function to select relevant data, enhancing model performance with fewer data points.

Al Surrogate Model:

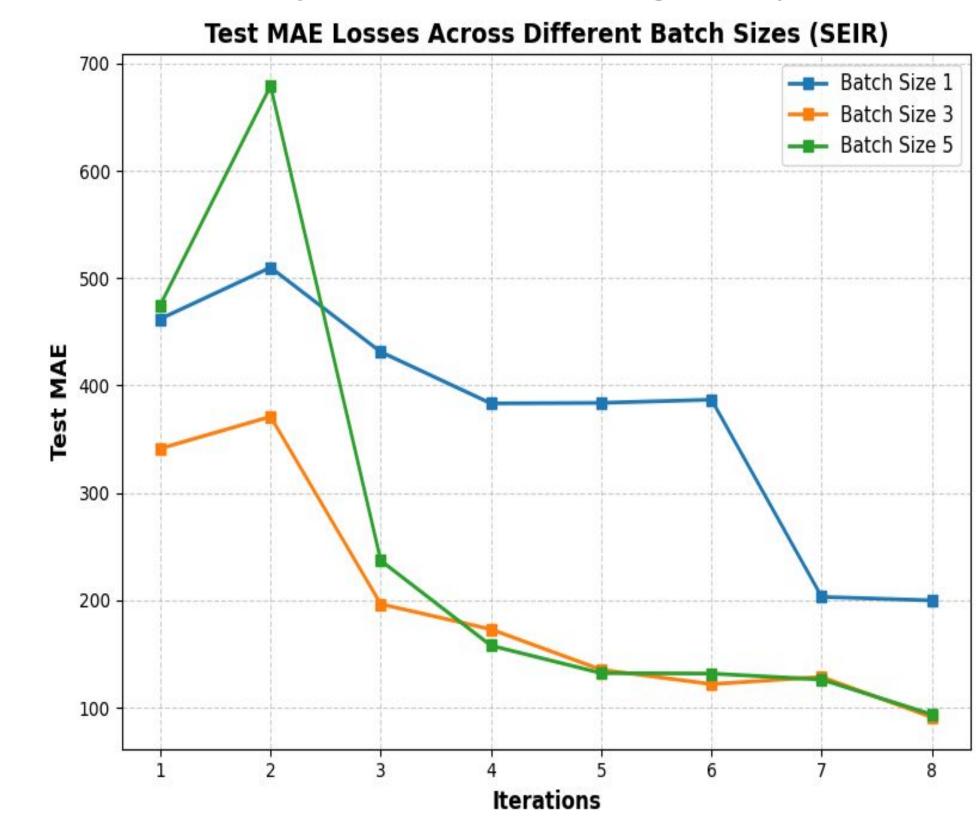
Al surrogate model uses BAL to optimize the amount of data needed to train down to 30% of data while achieving the same performance.

API Architecture Program



Results

SEIR training batch size experiment: Used greedy batch selection



Performance comparison between Leam-US and offline



Conclusion

Our API seamlessly integrates AI surrogate models with the GLEAM simulator, facilitating real-time epidemic forecasting. By employing Bayesian Active Learning, the model significantly reduces computational costs while preserving high accuracy.

References & QR Code

Balcan, D., B. Gonçalves, H. Hu, J. J. Ramasco, V. Colizza, and A. Vespignani. 2010. "Modeling the spatial spread of infectious diseases: The global epidemic and mobility computational model." Journal of Computational Science 1(3): 132-145.

Petrillo, F., P. Merle, N. Moha, and Y.-G. Guéhéneuc. 2016. "ARE REST APIs for cloudcomputing well-designed? An exploratory study." In Lecture Notes in Computer Science.:157-170.

Wu, D., R. Niu, M. Chinazzi, Y. Ma, and R. Yu. 2023. "Disentangled multi-fidelity deep Bayesian active learning." In International Conference on Machine Learning. PMLR

