Applications of Exascale Agent-based Modelling: Predicting Migration Patterns and Infectious Diseases

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Content

- Agent-based simulation applications:
 - Conflict-driven forced displacement (Flee)
 - Disaster-induced displacement (DFlee)
 - Flu and Coronavirus Simulator (FACS)
- Three-pronged attack to achieve exascale
- Summary

Modelling conflict-driven displacement

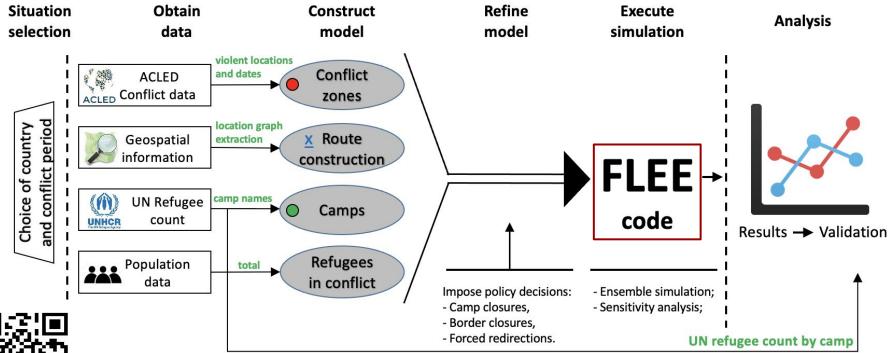
Motivation:

- Conflict erupts, people flee
 - Where do they go?
- Can predicting their arrival help organisations to effectively allocate support in advance?
- How do humanitarian decisions affect them?
 - Border closure.
 - Camp placement.
- Better understand historical processes.
- Inform decision-making, and possibly public awareness.



1st paper: Groen, D., 2016. Simulating refugee movements: Where would you go?. *Procedia Computer Science, 80*, pp.2251-2255.

Flee Simulation Development Approach





Suleimenova, D., Bell, D. & Groen, D. A generalized simulation development approach for predicting refugee destinations. Sci Rep 7, 13377 (2017)

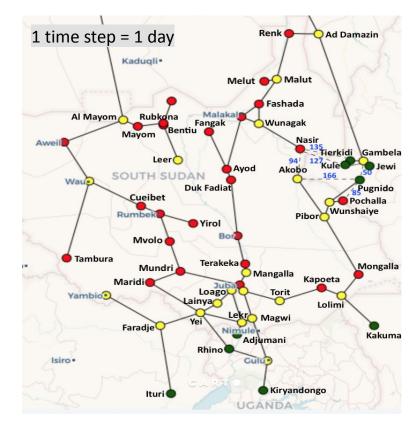
Introducing Flee

- Agent-based model for forecasting conflict-driven displacement.
- Predicts where people may go, given a developing conflict.
- Existing models for 13 historical conflicts offer a starting point.
- Open source, explicit assumptions.
- <u>https://flee.readthedocs.io</u>

Initial Code: Suleimenova, D., Bell, D. & Groen, D. A generalized simulation development approach for predicting refugee destinations. *Sci Rep 7*, 13377 (2017).

Forecasting challenge: Groen, D., Suleimenova, D., Jahani, A. and Xue, Y. Facilitating simulation development for global challenge response and anticipation in a timely way. Journal of Computational Science, 72, p.102-107 (2023).

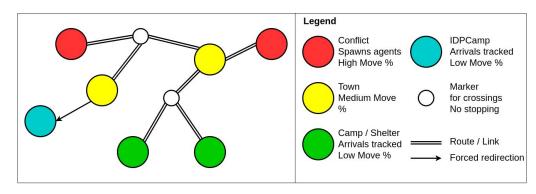
*#**FLEE



Basics of the logic



- Each agent = 1 displaced person.
 - Placed in a conflict zone, will move around in search of a camp.
- Each agent decides:
 - Do I stay put, or move to a neighbouring location?
 - If I move, which location shall I go to?
- Factors such as distance, perceived safety, ethnic match, distance from home can be introduced to shape the decisions.



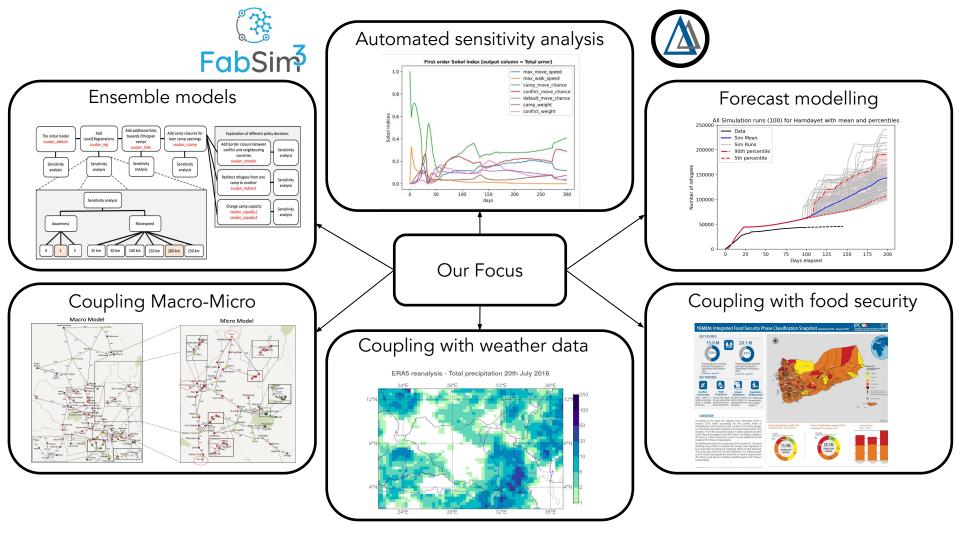
Main simulations so far





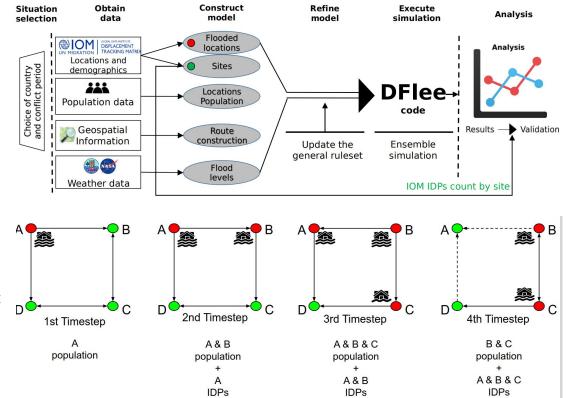


We validated our approach against refugee registration data from UNHCR, and we able to predict >75% of the arrivals correctly across four conflicts.



Disaster-induced displacement

- We investigate the movement of IDPs during a flood event in small regions rather than large regions and countries.
- Despite similarities between conflicts and disasters, there are some differences in terms of simulation settings, ruleset and destinations between the two events.
- An extension of the Flee code.
- Considering health facilities in route selection decision-making is very important as many of such facilities can temporarily shelter the evacuated people.



Flu and Coronavirus Simulator (FACS)

• Decision-makers need reliable and reproducible forecasts of the spread of COVID-19 in local regions.

So we established a local collaboration, with volunteers in the Department combined with some redirected effort from the HiDALGO EU Horizon project.

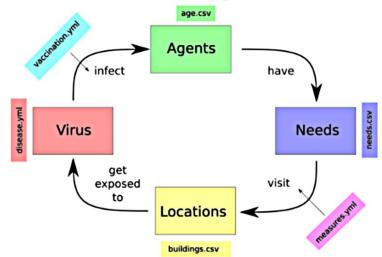
Our aim is to support:

- different lockdowns and other interventions,
- different viral strains,
- and later: different vaccination efficacies and strategies.

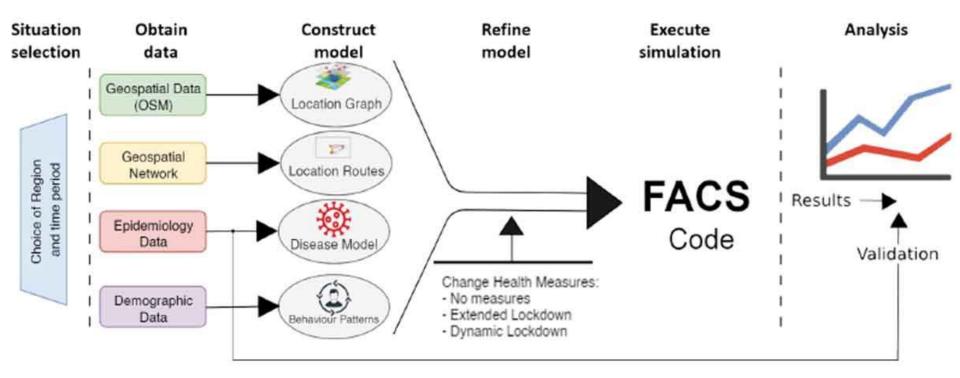
... on the hyperlocal sub-national level.

https://facs.readthedocs.io





FACS Simulation Development Approach



FACS code

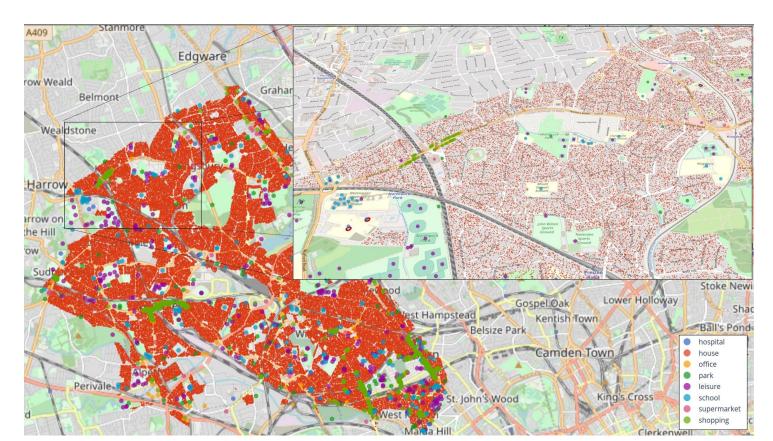
- Written in Python 3.
- Single runs support up to 1M households.
- Runtime: 1-8 hours
- FACS can efficiently run in parallel, using up to ~128 cores.
- Applications:
 - W. London NHS Trusts (13 forecasting reports),
 - UK Trial with Health Security Agency,
 - Lithuania, Turkey, Romania.

Dependency graph			
Dependencies Dependents Dependabot			
These dependencies are defined in facs's manifest files, such as requirements.txt and docs/requirements.txt.			
Dependencies defined in requirements.txt §			
> 🕙 matplotlib / matplotlib			
numpy / nu			
> 🛞 pandas-dev / pandas			
> m plothy / plothy.py plothy			
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Example input: Location graph



Measures: Non-pharmaceutical Interventions

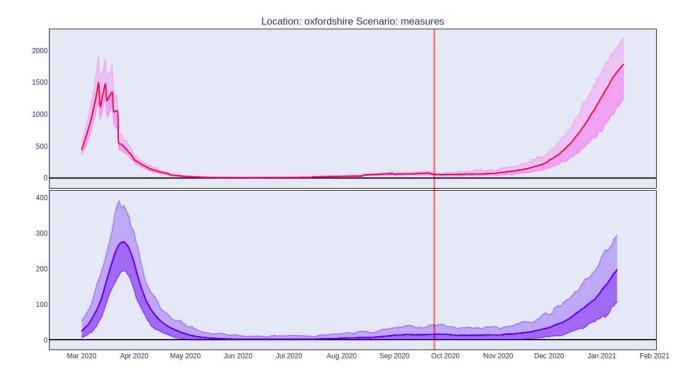
• Timeline of lockdown measures and closures provided in a yaml file

Table 1. Sur	mmary of locke	own measure	s implemented	during the simulation.

Date	Days since start of simulation	Lockdown Measures
March 01, 2020	0	Start of simulation
March 21, 2020	20	First lockdown
April 22, 2020	52	Peak of lockdown measures
September 01, 2020	184	Schools re-opening
November 05, 2020	249	Second lockdown
December 02, 2020	276	Restrictions lifted
December 23, 2020	297	Christmas bubble
January 06, 2021	311	Third lockdown
March 08, 2021	372	Restrictions lifting
April 04, 2021	399	End of simulation

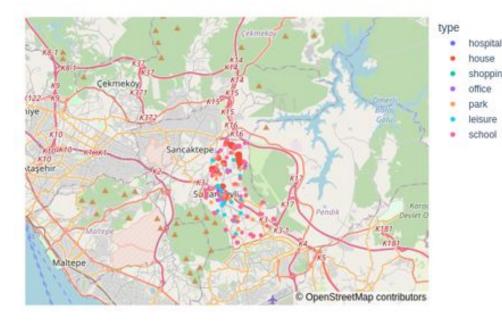
1/3/2020:
traffic_multiplier: 0.8
12/3/2020:
work_from_home: 0.1
social_distance: 0.1
traffic_multiplier: 0.5
16/3/2020:
partial_closure:
leisure: 0.3
work_from_home: 0.325 #assumption is that uptake between 11 and 21 Mar is gradual
social_distance: 0.25 #assumption is that uptake between 11 and 21 Mar is gradual
mask_uptake: 0.05
traffic_multiplier: 0.4
20/3/2020:
closure: ["leisure"]
work_from_home: 0.45 #assumption is that uptake between 11 and 21 Mar is gradual
social_distance: 0.5 #assumption is that uptake between 11 and 21 Mar is gradual
mask uptake: 0.05
traffic_multiplier: 0.3
23/3/2020:
partial_closure:
school: 1.0
shopping: 0.7 #indicates full school closure, but partial_closure is used to allow keyworkers to be added.
closure: ["leisure"]
social_distance: 0.75
mask_uptake: 0.05
mask_uptake_shopping: 0.1 work_from_home: 1.0
traffic multiplier: 0.25
2/4/2020:
partial closure:
school: 1.0
schopting: 0.8
closure: ["leisure"]
ctosure. [tersure]

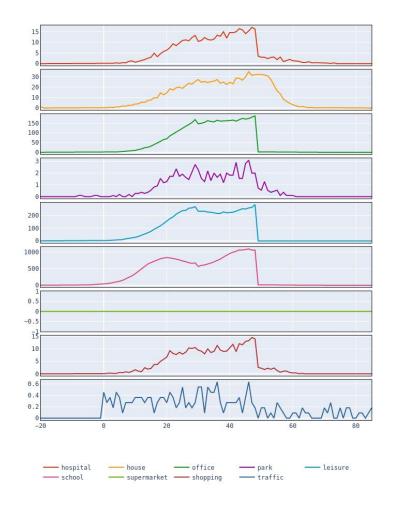
Example result: Oxfordshire Tier-2 (100 run ensemble)



Granular Details of Infections

- Building types where most infections take place.
- Geographical hot-spots as a function of time.





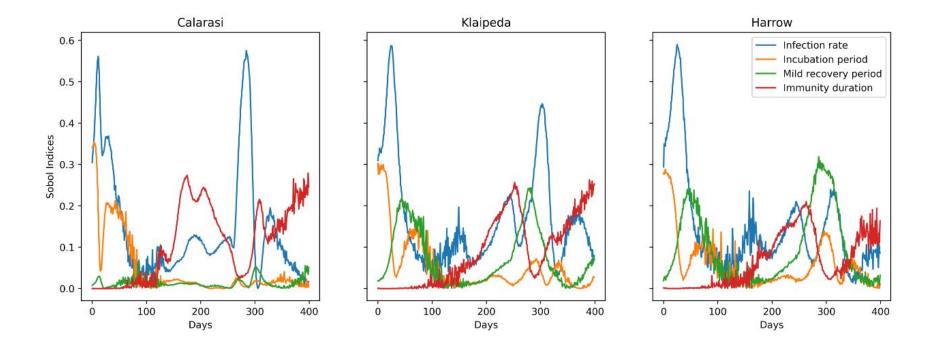
shopping

office

park

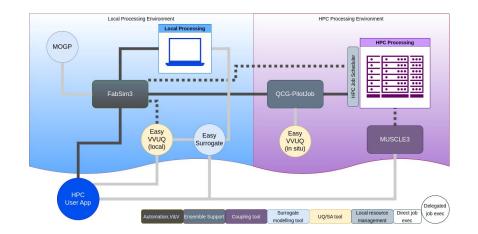
leisure

Sensitivity Analysis: Impact of Geography on



Need for Exascale Computing in FACS

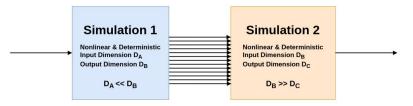
- Sensitivity Analysis
 - Disease parameters
 - Demographic characteristics
 - Interventions
- Multi-scalability
 - 'Large-picture' analysis nation-wide predictions
 - Detailed granularity admissions in a particular hospital
- Coupling with other simulations
 - UM Impact of climate on infectious diseases
 - CHARM Modifications to hospital bed management



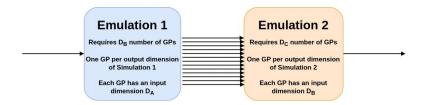
Need for Exascale Computing in FACS

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Given: Linked Simulation



Target: Linked Emulation



Three-pronged attack to achieve exascale

• Parallelizing/enhancing single scale models.

More cores, faster time to completion.

• Constructing multi-scale / coupled models.

More cores, more couplings, more simulated phenomena incorporated.

Adding VVUQ + Many-Objective Optimisation

More cores, more robustness, reproducibility and reliability.

Summary

- Active modelling efforts:
 - Flee, DFlee and FACS



- Improving the quality of our models by validating settings.
- Search for optimal configurations to model different scenarios.
- Sensitivity analysis: testing which assumptions matter most.
- Collaborations and projects









Upcoming events

□ The ExCALIBUR Workshop on Exascale Computing @ ICCS.

- <u>https://excalibur.ac.uk/events/iccs-excalibur-day</u>
- □ 1 July 2024, Malaga, Spain.
- **G** Register for the ExCALIBUR Workshop via the ICCS EasyChair registration process.



Questions?

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