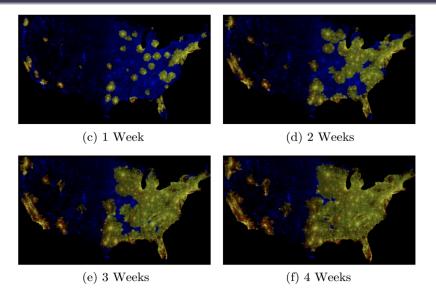
Case study: an epidemiological model

eX Modelo Summer School

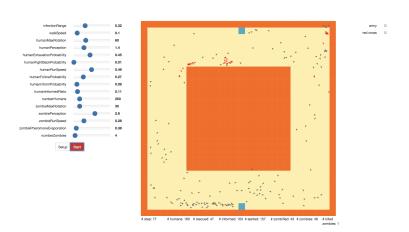
May 31st 2021



Simulation of the 2010 Zombie outbreak in the US [Alemi et al., 2015]



- ▶ 2007: first outbreak in Iceland, relatively contained through ad-hoc measures ("no-zombie" policy)
- ▶ 2010: it becomes pandemic
- ▶ 2010-2015: no clear records of events
- 2015-2018: reorganization of institutions, the MOLE (Medical Overview of Ludicrous Experiments) center in Chongqing gathers observational data from many local invasions across the world
- 2019: they released the first version of the model ZOMBIE (Zone of Optimal Management for Bacillus Infecting Everyone) and successfully applied
- ▶ June 2019: the first eXModelo studied the model and began saving the world
- September 2020: the second eXModelo saved the world a bit more but a pandemic messed everything up in the meantime



Local scale agent-based model

Let's get your hands on it



- ➤ A submodel is available at https://om.exmodelo.org/coop. Try the GUI and changing parameters
- Most of next courses will be based on that model (additional processes will be detailed when needed)



- Simulate agent-level collective movements at the scale of a district
- ▶ Include behavioral processes for human (panic, search for rescues, ...) and zombies (self-organization, spontaneous attacks, ...), which can be adapted to local settings
- Include realistic pedestrian dynamics and realistic spatial configuration, which can be applied to local configuration

Objective of the model: better understand the Zombie invasion process; possible applications to optimal policies and behavioral prevention to minimize the impact of recurring invasions

Issue with model application: model has many parameters and processes, model behavior is unknown, application may be strongly case-dependent

 \rightarrow we need **YOU** to understand this model to save the world



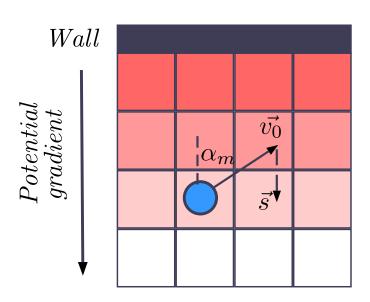
- Humans and Zombies walk/run randomly (smoothed random walk) in an open urban space (movement parameters: rotation angle, walk and run speed)
- ► Interactions: human flee from zombie, zombies run for food, fight when encounter
- Humans can be rescued and information on the existence of rescues propagates between humans
- Additional processes in a multi-modeling approach (army, vaccination, . . .)



Multiple approaches to pedestrian simulations:

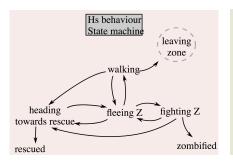
- ▶ Social force models [Helbing and Molnar, 1995]
- Granular flows [Cristiani et al., 2011]
- Behavioral models [Antonini et al., 2006]
- Cellular automatons [Burstedde et al., 2001]
- Potential field [Jian et al., 2014]

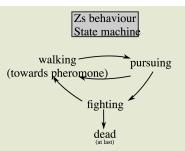
The ZOMBIE model takes the last approach, relatively realistic in a panic setting



Agents state machines









- Some spots allows informed humans to be rescued and get out of the world
- ► An initial ratio of humans humanInformRatio is informed of the existence of rescues
- Informed humans which are not in a panicking state follow a specific potential field leading to rescues
- A human can inform an other one at the same location with a probability humanInformProbability

With the additional parameter humanFollowProbability (probability for a human to begin running and follow when they encounter an other running human), the submodel with three parameters is aimed at studying cooperation between humans.





quarantine:
no rescue zones.
You'd better run.



jaude: one rescue zone of width 1

Alternative worlds . . . and many more automatically generated (spatial sensitivity module on Wednesday)

A flexible and more general model



- ► The general model, of which the three-parameter model is a particular case has more processes (deactivated or fixed)
- No need to know them for now, with the aim to understand the submodel on which we have the less knowledge
- ▶ Multi-modeling and model embedding (model as black box)

The model in practice



- Model implemented in scala (highly performant language mixing the best from functional programming and object programming)
- Simple GUI for interactive exploration
- Integrated into OpenMOLE as a jar plugin in the GUI



Scala syntax to run an execution of the model

```
import zombies._
val rng = new scala.util.Random(replication)

val result = zombieInvasion(
    humanFollowProbability = humanFollowProbability,
    humanInformedRatio = humanInformedRatio,
    humanInformProbability = humanInformProbability,
    zombies = 4,
    humans = 250,
    steps = 500,
    random = rng
)

val humansDynamic = result.humansDynamic(20)
val zombiesDynamic = result.zombiesDynamic(20)
```

References I



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