







INVESTICE DO ROZVOJE VZDĚLÁVÁNÍ

Agent-based Modeling and Simulation

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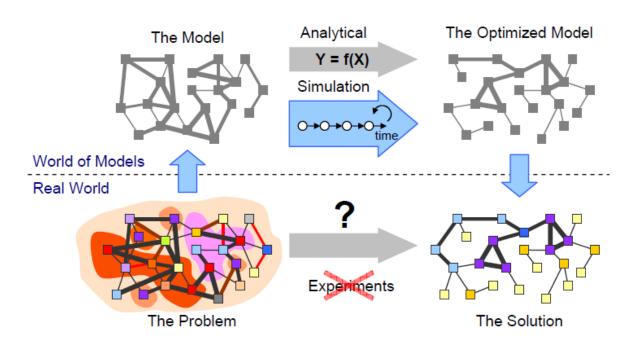
Introduction





Introduction

Simulation is a key tool for obtaining **insight** and **foresight** regarding the operation of complicated systems

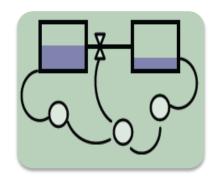


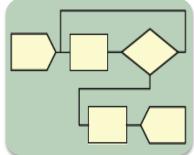
From: Borshchev, A. et al (2004): From system dynamics and discrete event to practical agent based modeling: Reasons, techniques, tools



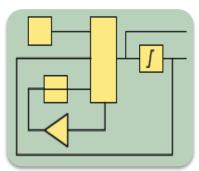


S&M Approaches









System dynamic

- states, feedbacks and delay structures
- continuous
- global, aggregate view

Discrete Event

- entities and resources
- discrete, eventbased
- global entity processing algorithm

Agent Based

- active entities and the environment
- decentralized, individual perspective
- global behavior emerges

Dynamic Systems

- state variables and differential equations
- direct physical meaning, no aggregation





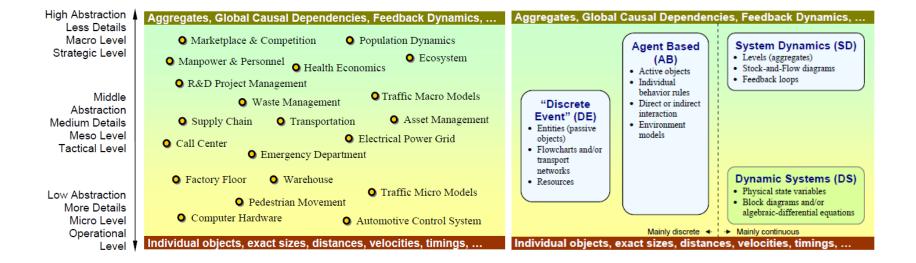








Levels of Abstraction



From: Borshchev, A. et al (2004): From system dynamics and discrete event to practical agent based modeling: Reasons, techniques, tools





Illustrative Example

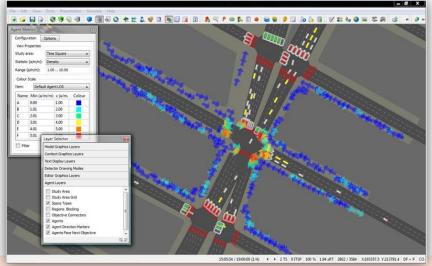
Pedestrian simulation

 Each pedestrian modeled as an agent sensing the environment and interacting with other pedestrian agents

The model allows to

- determine crowd flows and densities under various scenarios
- optimize crowded public spaces for capacity, comfort and safety









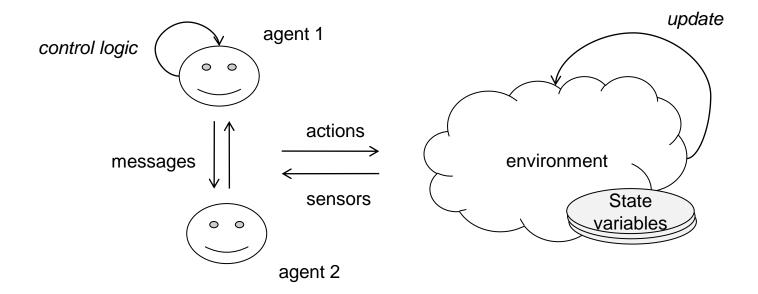




Architecture of the Agent-based Simulation



Structure of the Agent-based Simulation



Agents drive the model through local behaviors and direct and indirect interaction with each other and with the environment

Environment state is modified by agent actions and/or agent-independent processes (e.g. weather)





Agent

Key component of an-agent based model

Proactive entity acting towards its objective

Key operations

- processing sensory inputs
- selecting actions to perform
- managing communication

Additional operations

- adaptation
- deliberation





Agent's Control Logic

BDI (belief-desire-intention) architecture as a foundation

Simple control logic implementations used in practice

- (hierarchical) finite state machines
- rule engines
- scripts

Calls to special-purpose subsystems

- navigation
- environment perception

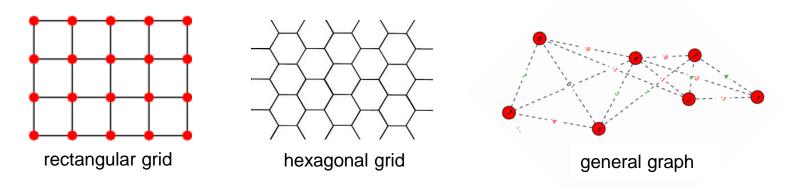




Environment

Shapes agent's operation, mediates inter-agent interaction, can impact inter-agent communication

Abstract environments



Realistic/VR: open landscape, urban, indoor



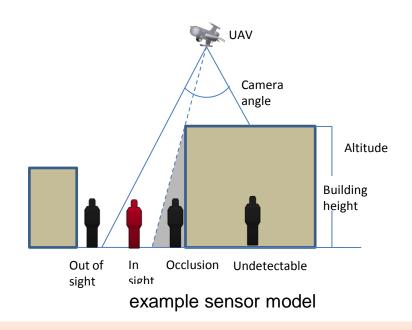
Sensors

Enables the agent to access the environmental state

- low-level direct perception (e.g. image from a camera)
- high-level interpreted scene (e.g. walls, people)

Efficient implementation crucial in more complex environments

- partitioning
- caching







Developing agent-based simulations



Development Cycle

Design

- identify agents and environmental features
- choose the right modeling abstraction critical

Implementation

- agent's control logic
- environmental features and dynamics

Calibration and Validation

- micro-level and macro-level
- real-world data-based (if possible)





Calibration and Validation

Micro-level

- parameter correspondence (e.g. size, speed, visual range, ...)
- behavior correspondence (reaction under different circumstances)

Macro-level

- compare outcome distributions
- comparison with real system traces (start from the same initial conditions)

Generally very difficult due to inherent complexity of the system





Platforms and Tools

General platforms still only in an early stage

- academic: RePast, NetLogo
- commercial: AnyLogic
- A-Globe Simulation

Special-purpose platforms more mature

- traffic modeling: <u>AIMSUN</u>
- pedestrian modeling: <u>LEGION</u>

GIS tools and data sources

- Google Earth, NASA WorldWind
- http://www.openstreetmap.org/





Advantages

Captures emergent phenomena

- the whole is not a sum of parts system cannot be abstracted
- non-linearity, discontinuity, phase-transitions

Provides a natural description of the system

- agents easy to identify in the real-world system
- agents' activities easier to describe than global processes

Flexible

- agents' behaviors can be easily extended/specialized
- size of the model can be scaled





When to Use

Complex interaction between entities

- interaction topology not fixed
- feedback relationships, network effects

Heterogeneity of entities

each agent has a different behavior

Behavior of individuals is complex

learning, adaptation

Unknown global process structure and/or parameters

Environment is crucial and agents are mobile





Application Areas

traffic and transport

optimization of traffic networks, understanding and eliminating congestion

pedestrians and crowds

capacity optimization, evacuation procedures

organizations

organization design optimization, operation risk estimation

markets and economies

computer networks

bandwidth usage estimation, worm infection modeling

security

crime modeling, vulnerability estimation





Conclusions





Conclusions

Most recent addition to modeling and simulation toolbox

Bottom-up approach

Most suitable for complex systems composed of autonomous, interacting entities

Allows high-fidelity models at the expense of high-computational cost

Mature tools exist for specific domains (e.g. transport, crowds)

General purpose platforms and tools still under development

