



ERASMUS+ KA2 Strategic Partnership  
2017-1-FI01-KA203-034721  
HELP – Healthcare Logistics Education and Learning Pathway



# Simulation in Healthcare



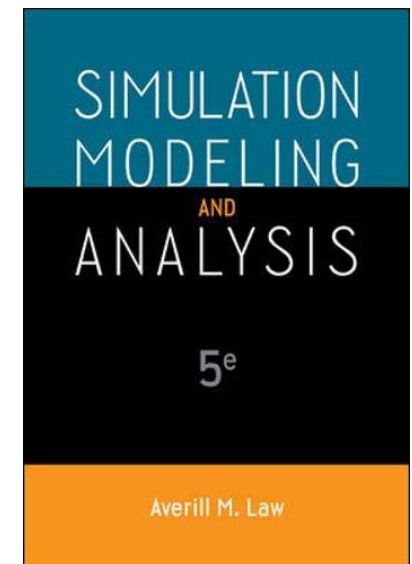


ERASMUS+ KA2 Strategic Partnership  
2017-1-FI01-KA203-034721  
HELP – Healthcare Logistics Education and Learning Pathway



- **Recommended reading**

Averill M. Law, Simulation modeling and analysis,  
5th edition





## What is simulation?

- *Wikipedia*: “Imitation of the operation of a real-world process or system over time. The act of simulating something first requires that a model be developed; this model represents the key characteristics or behaviors/functions of the selected physical or abstract system or process”
- Applies in many industries and fields
- Very popular and powerful method





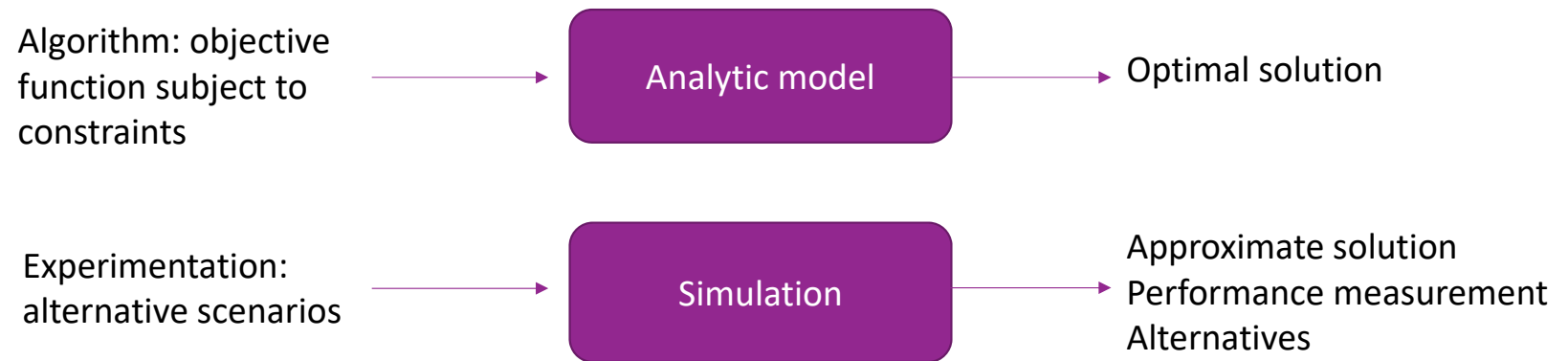
# What is simulation?

Simulation is a numerical technique for conducting experiments with certain types of logical or mathematical models, describing the behaviour of complex systems on a digital computer over extended periods of time

## Simulation is ...

- A **numerical** technique
  - Approximate solutions: solving mathematical problems is too complex to provide analytical solutions (e.g. behaviour of nonlinear systems)
  - Queueing theory vs. Simulation
    - Queueing theory: mathematical models to simulate waiting lines
      - If demand > capacity: models for optimizing capacity (trade-off capacity cost and service time)
      - Optimization is better; but complex reality with dynamic effects and random events  
→ seldom validly represented by analytical model due to unrealistic assumptions → simulation
- For conducting **experiments**
  - “what-if” analysis: evaluate various strategies (within the limits imposed by a criterion or set of criteria)

- Simulation vs. analytic models



## Simulation is ...

- With certain types of logical or mathematical **models**
  - Model = set of assumptions about how the system works
  - Study behavior of model rather than real system: easy, fast, cheap, safe
    - Model validation: relationship between real system and conceptual model (building the right model?)
    - Sensitivity analysis: input parameter modifications, impact assumption violations
- Describing the behaviour of complex **systems**
  - System = actual or planned business process (e.g. warehouse, production facility, hospitals)
    - Change of system state = event (e.g. number of doctors available in ED)
  - Learn about behaviour/performance of systems: design, measure, improve and control

## Simulation is ...

- On a digital **computer**
  - Computer software: Arena, Matlab, Enterprise Dynamics, AnyLogic, etc.
  - Verification: translate conceptual model to operational model in software language (building the model right?)
- Over extended periods of **time**
  - Dynamic
  - Simulation run much faster than real time
  - Reduce uncertainty of events in a time period (e.g. arrival pattern of patients)
    - Uncertainty represented by random number generator (distribution function)



# Why simulation?

## Advantages

- Complexity
- System variability
- Flexibility: changes in system
- Experimentation: reporting functionality (e.g. graphs)
- Simulation speed
- Visualization power (2D/3D)
- Cost-effective
- User-friendly



## Disadvantages

- Approximate solution: estimates
- Data requirements
- Validation and verification
- Time effort vs. Model accuracy (scope + level of detail)
- Learning how to simulate
- Computational feasibility

## When to use simulation?

- Analysis of business processes
  - Support decision-making by evaluating improvement strategies in a cost-effective, non-disruptive manner
  - Predict the system's performance when time evolves (dynamic)

### **NOT** useful if:

- Simple to calculate analytically
  - No stochastic (deterministic)

### **Healthcare** engineering solutions:

- Engineers (experimentation) vs. Doctors (evidence-based)
- Visualization power: understanding > awareness > commitment > impact

## Types of simulation

- Deterministic vs. Stochastic
  - Predictable system
  - Statistical distribution functions
- Discrete vs. Continuous
  - Event-based: change of system state at given points in time
  - State variables change every moment
- Static vs. Dynamic
  - Time independent decisions
  - Time dependent decisions

# Types of simulation

	Monte Carlo/ Markov queueing	Discrete-Event simulation	System Dynamics	Agent-based simulation
Static / Dynamic	S	D	D	D
Discrete / Continuous	D	D	C	C
Deterministic / Stochastic	S	S	D	S

# Types of simulation

## Monte Carlo simulation

- Mathematical technique:
  - Randomly generate inputs from probability distribution
  - No time dimension
- Example:
  - Patient arrival time and service duration are determined by flipping a coin (heads = 1 patient arrives in an hour/2 hours service time; tails = no patient arrives in an hour/1 hour service time)



ERASMUS+ KA2 Strategic Partnership  
2017-1-FI01-KA203-034721  
HELP – Healthcare Logistics Education and Learning Pathway



Time interval	Patient arrival	Queue	Service duration	Doctor consultation	Patient departure
8:00 – 8:59	Heads: #1	/	Heads	#1	/
9:00 – 9:59	Heads: #2	#2	Tails	#1	#1
10:00 – 10:59	Heads: #3	#3	Tails	#2	#2
11:00 – 11:59	Tails: /	/	/	#3	#3
12:00 – 12:59	Heads: #4	/	Heads	#4	/
13:00 – 13:59	Heads: #5	#5	Heads	#4	#4
14:00 – 14:59	Tails: /	/	/	#5	/
15:00 – 15:59	Heads: #6	#6	Tails	#5	#5
		<b>Total waiting time</b>	<b>Total service time</b>		<b>Total time in system</b>
		4	9		2+2+2+2+3+2=13



# Types of simulation

## Discrete-event simulation

- Model series of events over time (no change in system between events) = system state
  - Event list + event controller (simulator)
- Simulation components:
  - Entities move around, change status
  - Resources assigned to entities (entities compete for people, equipment, space)
  - Attributes (global/local characteristics of model/entities: parts in system, color/priority)
  - Queues
  - Statistical counters: measure performance indicators (waiting time, utilization rate, etc.)
- Micro-modelling (low abstraction level): operational and tactical decision-making

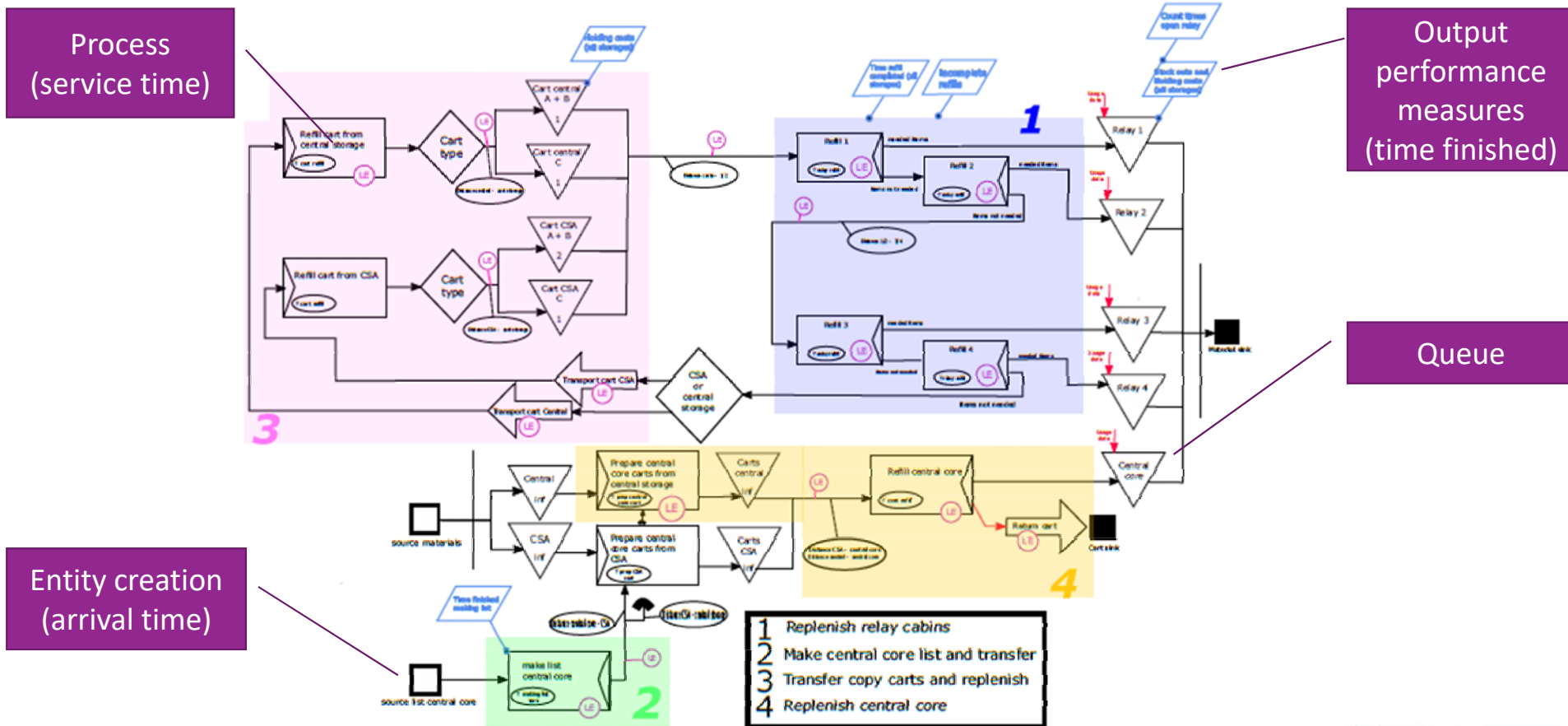
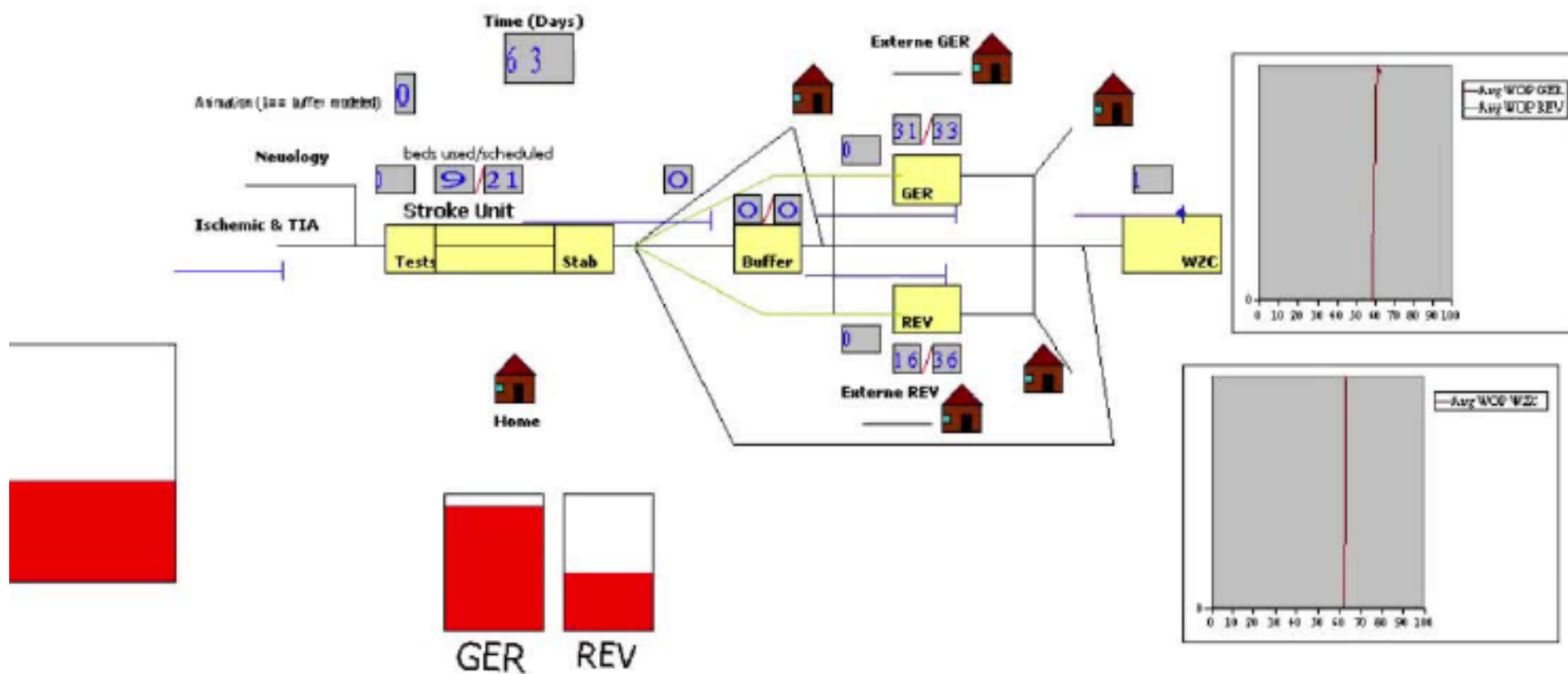


Figure 4.7: As-is scenario conceptual model





# Types of simulation

## System dynamics

- Abstract macro-modeling for strategic decision-making
  - Aggregate level vs. individual objects (DES)
  - Relationship between system elements using feedback loops
    - Stocks = state variables
    - Flow = rate (change value of stocks)
    - Loop when changes in a stock affect flows in/out the stock



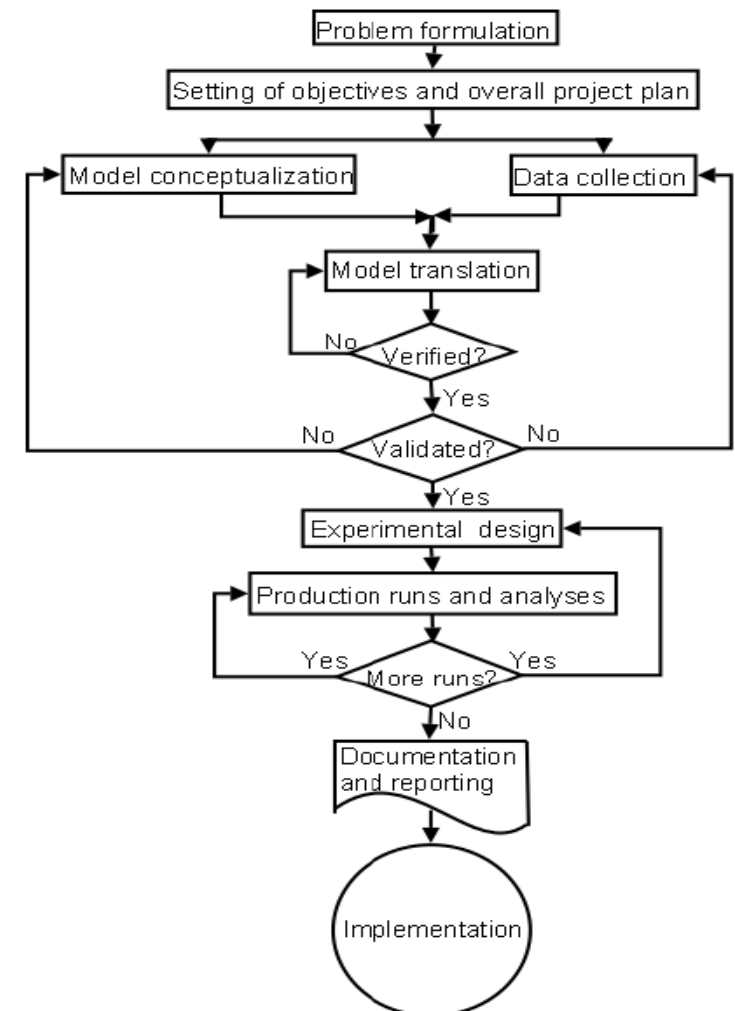
# Types of simulation

## Agent-based simulation

- Modeling of operations and interactions of autonomous agents to evaluate impact on system
  - Agent = active, autonomous entity with own goals and behaviours
  - System = interacting agents
    - No controller for how system operates

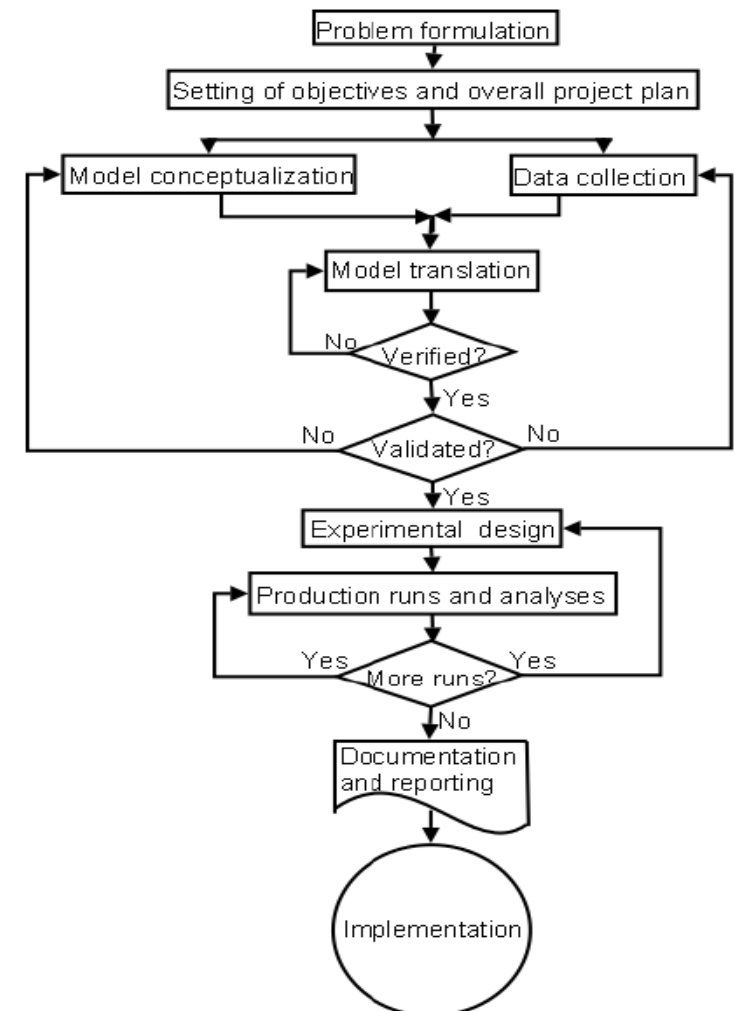
# Simulation study roadmap

- Problem definition
  - Understand problem situation
- Conceptual model (blueprint):
  - Scope (what to model?): modeling objectives, resources, input, output
  - Level of detail (how to model?): assumptions, simplifications
  - Data requirements
  - Object flow diagram: model constructs



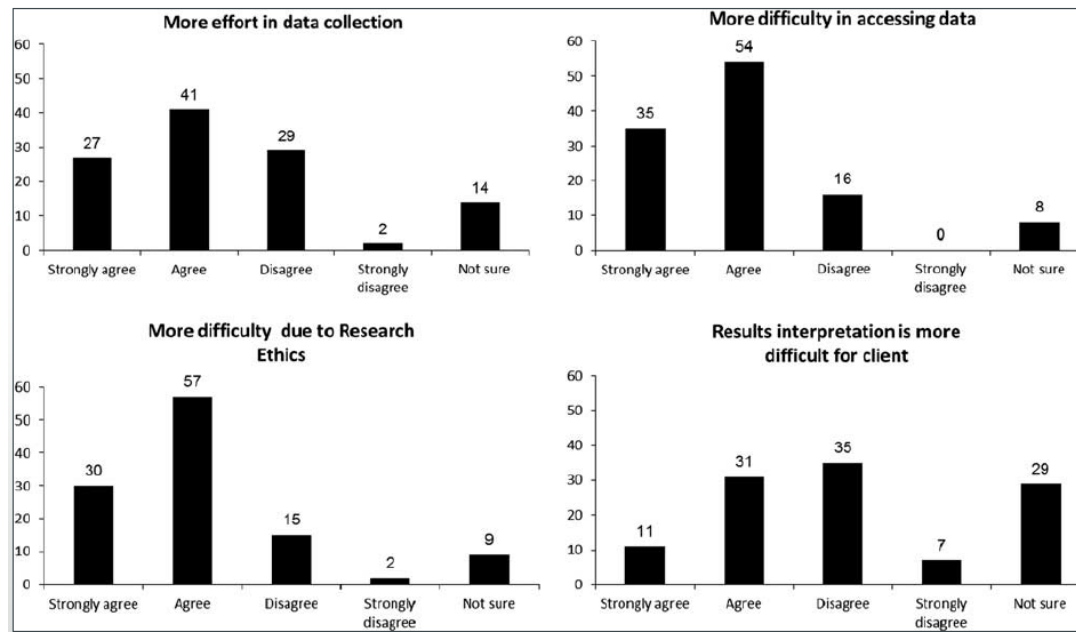
# Simulation study roadmap

- Model translation:
  - Computer-specific software
  - Verification and validation
    - Build the model right: debugging
    - Build the right model: sensitivity analysis
- Experimental design
  - Scenario analysis
- Replication
  - The more runs, the more accurate the performance measures
  - Confidence intervals, statistical t-test
- Reporting: analyze output data
- Implementation



# Simulation in healthcare

- Modeling healthcare systems is more difficult than in other sectors



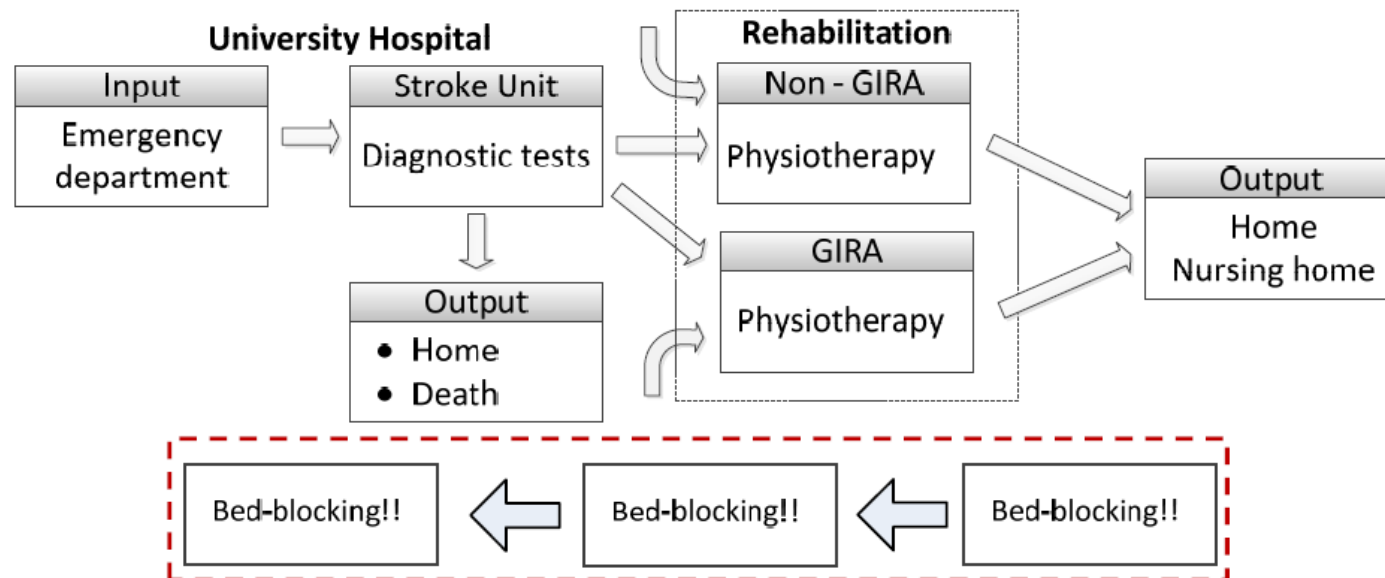
# Simulation as a tool to support healthcare decision making

- Successful application in manufacturing and business application, but rather novel in healthcare sector
- Healthcare organizations are subject to economic constraints and service-based targets  
→ Improve operational efficiency while maintaining quality of care
- DES as supporting tool for decision makers
  - Define operations, map processes and gathering data in structured way to understand current situation and identify improvement opportunities
  - Stimulate stakeholder commitment: trust in findings and ensure objective, data-driven decision-making
- Lack of implementation of results
  - As-Is vs To-Be comparison highlights importance of pairing DES with monitoring KPIs of improvement strategies
  - DES is a valuable tool to target operational improvement actions by managing variability associated with workflows

# Case studies – Patient logistics

## Simulation modeling for stroke patient flow

- Bed-blocking problem → patient occupies bed resource though medically ready





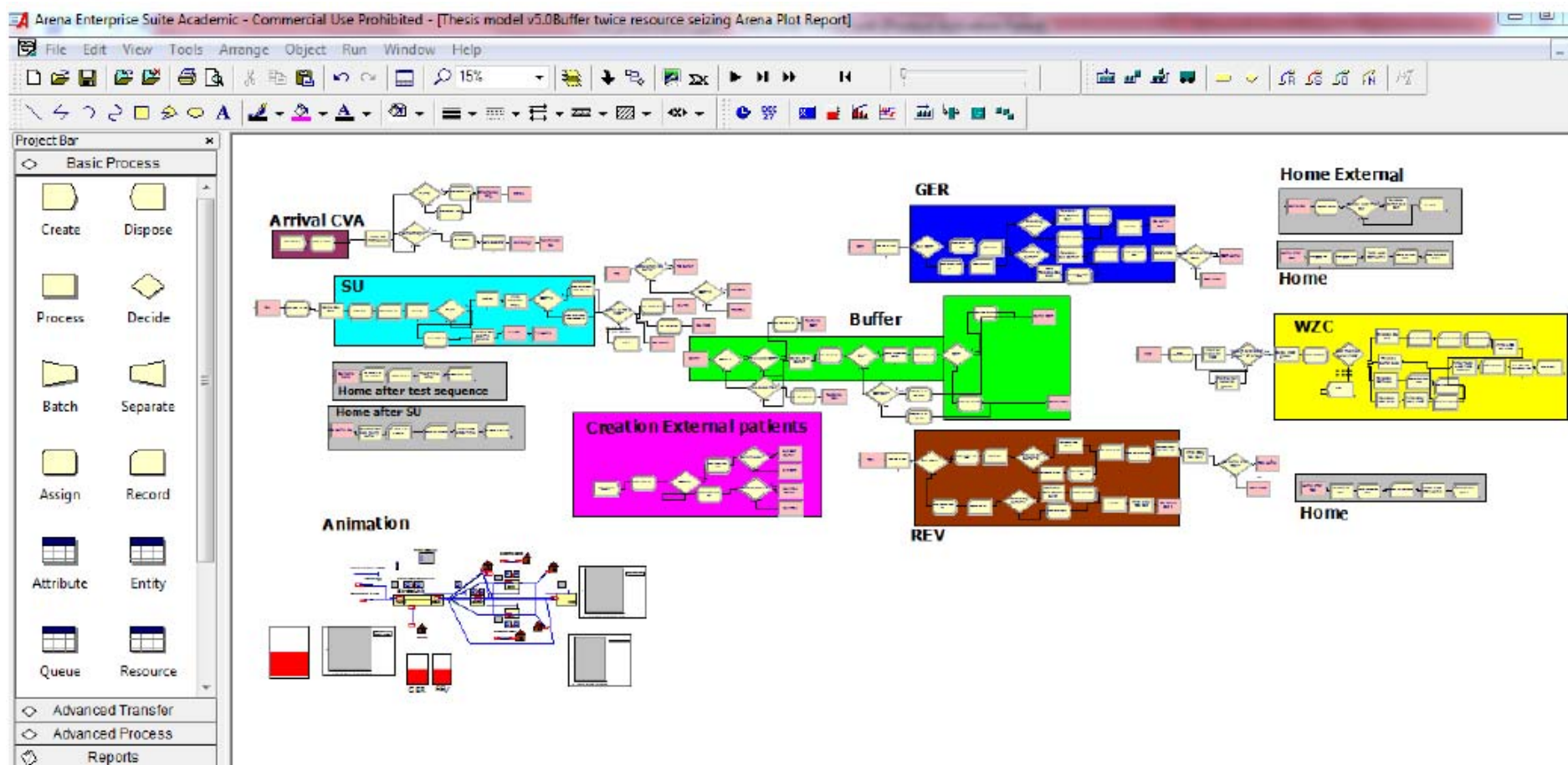


- Challenges:

- Bed-blocking increases patient waiting times for stroke treatment and rehabilitations
  - ➔ Diminished quality of healthcare
- Inefficient resource allocation/optimization with the integrated stroke patient care pathway
  - ➔ Balancing resources (e.g. MRI, CT scan, neurologist, nursing, etc.) with patient demand not straightforward

- Simulation roadmap

1. Map patient transfer process and resources along care pathway
  - Mapping tools: interview, brainstorming, patient records, observation, etc.
2. Data requirements
  - Patient flow statistics: patient arrivals per time, number of treatment sessions, patient waiting time, number of doctors, etc.
    - Distribution functions
3. Assumptions
  - Transportation means available when required
4. Simulation model
  - Patient arrival > diagnosis/treatment > transfer to rehabilitation center > rehabilitation > patient discharge





## 5. Model validation

- Comparison with patient data

## 6. Experimentation

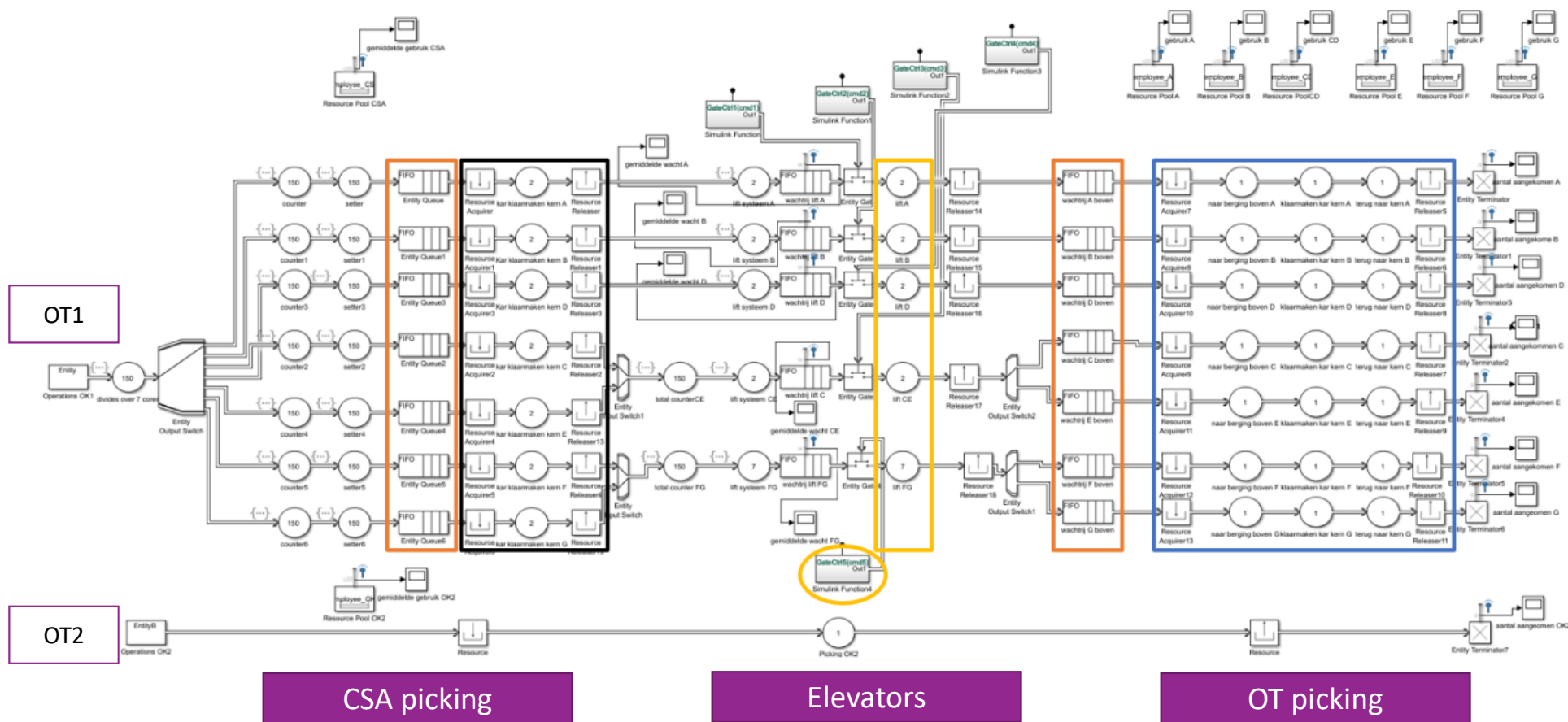
- Evaluate alternative improvement scenarios: determine optimal bed size to minimize patient length of stay
  - Unlimited bed capacity:
  - Varying transfer schedule
  - Vary age limit to geriatric ward
  - Implement shared ward



## Case studies – Materials logistics

### **Simulation modeling of surgical case cart distribution flow**

1. Variability in workflows between operating room clusters and reduce waste
2. Data: picking time, traveling time, items per case cart
3. Case carts assumed to be available when required
4. Simulation model





5. Debugging, use expert opinions (practical experience), structured walkthrough, run simplified model (predictable), observe animation
6. Experimentation:
  - Impact of standardization: streamline workflows
  - Impact of centralization: reduce logistics movements close to operating room
  - Impact of elevator: reduce idle time



**ERASMUS+ KA2 Strategic Partnership**  
**2017-1-FI01-KA203-034721**  
**HELP – Healthcare Logistics Education and Learning Pathway**



# Questions?






<p>This project has been funded with support from the European Commission.</p> <p>The European Commission support for the production of this publication does not constitute endorsement of the contents which reflects the views only of the authors, and the Commission cannot be held responsible for any use which may be made of the information contained therein.</p>
