Aachen Summer Simulation Seminar 2014

Lecture 03 Introduction to Conceptual Modelling

Peer-Olaf Siebers



pos@cs.nott.ac.uk

Motivation

- Define what a conceptual model is and how to communicate such a model
- Demonstrate how to develop a conceptual model





ASSS 2014

Introduction



- Importance of conceptual modelling (or model design)
 - The modeller along with the clients determines the appropriate scope and level of detail to model, a process known as conceptual modelling
 - Model design impacts all aspects of the study
 - A high proportion of the benefits of a simulation study is obtained just from the development of the conceptual model
 - For the development of the conceptual model we often seek answers to questions that have not previously been ask
 - Effective conceptual modelling may even lead to the identification of a suitable solution without the need for any further simulation work



Introduction



- What about the following argument:
 - The emergence of modern simulation software has reduced or even removed the need for conceptual modelling?
 - The modeller can move straight from developing an understanding of the real world problem to creating a computer model
 - The software allows rapid model development and prototyping but it does not reduce the level of decision making about the model design
- What about the following argument:
 - Power and memory of modern hardware and the potential of distributed software has increased the need for conceptual modelling?
 - Increase in complexity of simulation models; modellers build more complex models because software/hardware allows them to do so
 - Models are being developed that are far more complex than they need to be; careful model design is increasing in importance



- Definition (Robinson 2008a):
 - The conceptual model is a *non-software specific description* of the computer simulation model (that will be, is or has been developed), describing the objectives, inputs, outputs, content, assumptions and simplifications of the model.
- Conceptual modelling is more an art than a science; therefore it is difficult to define methods and procedures





- Key components of a conceptual model:
 - **Objectives:** The purpose of the model
 - Inputs: Elements of the model that can be altered
 - **Outputs:** Measures to report the results from the simulation runs
 - *Content:* Components represented in the model and their interconnections
 - Assumptions: Uncertainties and believes about the real world to be incorporated into the model
 - *Simplifications:* Reduction of the complexity of the model





- Basic conceptual model for booking clerk @ theatre:
 - **Objectives:** Serve 95% of customers in less than 10 minutes
 - Inputs: Arrival rates, service rates, number of clerks
 - Outputs: % of customers queuing for less than 10 minutes; histogram of waiting time for each customer in the queue; clerk utilisation
 - *Content:* Personal enquirers; phone callers; inter arrival time distribution; service time distribution; queuing priority
 - Assumption: Unlimited queues (we do not know space availability)
 - *Simplifications:* Queuing discipline (no jockeying, balking, leaving)
- Remember:
 - Assumptions are a facet of limited knowledge or presumptions
 - Simplifications are a facet of the desire to create simple models



ASSS 2014





- Requirements of a conceptual model:
 - Validity
 - Credibility
 - Utility
 - Feasibility
- What do these terms mean?



- Requirements of a conceptual model (Robinson 2004):
 - Validity: A perception, on behalf of the modeller, that the conceptual model will lead to a simulation model that is sufficiently accurate for the purpose at hand
 - Credibility: A perception, on behalf of the clients, that the conceptual model will lead to a simulation model that is sufficiently accurate for the purpose at hand
 - Utility: A perception, on behalf of modeller and clients, that the conceptual model will lead to a simulation model that is useful as an aid to decision making within the specified context
 - Feasibility: A perception, on behalf of modeller and clients, that the conceptual model will lead to a simulation model



Model complexity and accuracy

- Aim: Keep the model as simple as possible to meet the objectives of the simulation study
- Advantages of simpler models:
 - They can be developed faster
 - They are more flexible
 - They require less data
 - They run faster
 - Results are easier to be interpreted





Model complexity and accuracy

- 80/20 Rule
 - 80 percent of accuracy is gained from only 20% of complexity; beyond this there is diminishing returns from increasing levels of complexity
 - Increasing the complexity (scope and level of detail) too far might even lead to a less accurate model since the data and information are not available to support the detail being modelled





ASSS 2014

Model complexity and accuracy

- 80/20 Rule
 - 80 percent of accuracy is gained from only 20% of complexity; beyond this there is diminishing returns from increasing levels of complexity
 - Increasing the complexity (scope and level of detail) too far might even lead to a less accurate model since the data and information are not available to support the detail being modelled
- It is important to consider both, constructive simplicity and transparency.
 - Constructive simplicity: Attribute of the model
 - Transparency: Attribute of the client



- Simplification entails reducing the scope and the level of detail in a conceptual model
 - Scope reduction: Removing components and interconnections that have little effect on model accuracy
 - Detail reduction: Representing more simple components and interconnections while maintaining a satisfactory level of model accuracy
- Remember:
 - Most effective approach to simplification is to start with the simplest model possible and gradually add to its scope and level of detail; once a point is reached in which the study objectives can be addressed, then no further details should be added





- Methods (scope or level of detail reduction?)
 - Aggregation of model components
 - Black box modelling
 - Grouping entities
 - Excluding components and details
 - Replacing components with random variables
 - Excluding infrequent events
 - Reducing the rule set
 - Splitting models



- Methods
 - Aggregation of model components [detail reduction]
 - Black box modelling
 - Grouping entities
 - Excluding components and details [scope reduction]
 - Replacing components with random variables [detail reduction]
 - Excluding infrequent events [scope reduction]
 - Reducing the rule set [detail reduction]
 - Splitting models [advantage: individual models run faster]
- Remember: Over-simplification can make a model less transparent and thereby reducing its credibility



- Representing the conceptual model (examples):
 - System Dynamics (SD)
 - Causal loop diagrams; stock and flow diagrams
 - Discrete Event Simulation (DES)
 - Component list; process flow diagram; logic flow diagram; activity cycle diagram; combining Petri net and UML static structure diagrams (Pels and Goossenaerts 2007); class diagram to support OO DES
 - Agent Based Simulation (ABS)
 - UML + AgentUML (class, component, sequence, deployment, state chart, use cases, and activity diagrams) (Bommel and Müller 2008); coloured Petri nets (Jensen et al 2007)



- Representing the conceptual model (examples):
 - System Dynamics (SD)
 - Causal loop diagrams; stock and flow diagrams
 - Discrete Event Simulation (DES)
 - Component list; *process flow diagram*; logic flow diagram; activity cycle diagram; combining Petri net and UML static structure diagrams (Pels and Goossenaerts 2007); *class diagram* to support OO DES
 - Agent Based Simulation (ABS)
 - UML + AgentUML (*class*, component, sequence, deployment, *state chart*, use cases, and activity *diagrams*) (Bommel and Müller 2008); coloured Petri nets (Jensen et al 2007)



• DES Example: M/M/1/n Queue

- A single server system with ...
 - A queue capacity of n
 - An infinite calling population
 - Poisson (random) arrival process (inter-arrival times are exponentially distributed) and service times are also exponentially distributed





• DES Example: M/M/1/n Queue

- A single server system with ...
 - A queue capacity of n
 - An infinite calling population
 - Poisson (random) arrival process (inter-arrival times are exponentially distributed) and service times are also exponentially distributed

| Component | Detail |
|-----------|--|
| Customers | Inter-arrival time (exponentially distributed) |
| Queue | Capacity |
| Service | Service time (exponentially distributed) |

Component list







UNITED KINGDOM · CHINA · MALAYSIA

- Framework for conceptual modelling
 - Four key elements
 - 1. Develop an understanding of the problem situation
 - 2. Determine the modelling objectives
 - 3. Design the conceptual model: Inputs and outputs
 - 4. Design the conceptual model: The model content
- Remember that conceptual modelling is an iterative process!
- For more examples see Robinson (2004) Appendix 1+2
 - Online version available from the library catalogue





- 1. Develop an understanding of the problem situation
 - Clients might not have a good understanding of the cause and effect relationships within the problem situation
 - Clients have different world view
 - While learning from clients the modeller needs to play an active role
 - Modeller needs to confirm his/her understanding by providing a description of the problem situation for the client
- Problem situation and understanding of it will both be changing during the simulation study





- Case Study: Fast-Food Restaurant (Robinson, 2004):
 - A fast-food restaurant is experiencing problems with one of its branches in its network. Customers regularly complain about the length of time they have to queue at the service counters.
 - It is apparent that this is not the result of shortages in food, but a shortage of service personnel.







ASSS 2014

- 2. Determining the modelling objectives
 - Modelling objectives determine the nature of the model
 - Modelling objectives determine level of abstraction and simplification
 - Modelling objectives are a reference point for model validation
 - Modelling objectives guide for experimentation
- The purpose of the modelling study is not the development of the model itself but to develop a tool to aid decision making
- Bad practice: Developing models that do not serve any useful purpose, e.g. models that are looking for a problem to solve



- 2. Determining the modelling objectives (cont.)
 - Forming the objectives:
 - By the end of the study what do we hope to achieve?
 - What does the client want to achieve?
 - What level of performance is required?
 - What constraints must the client (modeller) work within?
 - Modeller should be willing to suggest additional objectives and to redefine or eliminate objectives suggested by the clients
 - It is important that the clients understands what a simulation model can and cannot do for them; managing the expectations of the client





- Case Study: Fast-Food Restaurant
 - What does the client want to achieve?
 - What level of performance is required?
 - What constraints must the client (modeller) work within?
- Objective:
 - The number of service staff required during each period of the day to ensure that 95% of customers queue for less than 3 minutes for service.
- Constraint:
 - Due to space constraints, a maximum of six service staff can be employed at any one time.



- 3. Design the conceptual model: Inputs and outputs
 - Experimental factors (inputs):
 - Often, they are the means by which it is proposed that the modelling objectives are to be achieved
 - They can be either qualitative or quantitative
 - They are often under control of the clients; however, also factors that are not under control of the client should be considered as this improves the understanding of the real system
- Remember: If possible, the range over which experimental factors are to be varied as well as the method of data entry should be defined



- 3. Design the conceptual model: Inputs and outputs (cont.)
 - Responses (outputs):
 - Measures used to identify whether the objectives have been achieves
 - Measures used to identify reasons for failure to meet objectives (e.g. bottlenecks)
- During the course of the simulation study review the experimental factors and responses when objectives are changing!







- Case Study: Fast-Food Restaurant
 - Objective:
 - The number of service staff required during each period of the day to ensure that 95% of customers queue for less than 3 minutes for service.
 - Constraint:
 - Due to space constraints, a maximum of six service staff can be employed at any one time.
 - Experimental factors?
 - Responses?





- Case Study: Fast-Food Restaurant
 - Objective:
 - The number of service staff required during each period of the day to ensure that 95% of customers queue for less than 3 minutes for service.
 - Constraint:
 - Due to space constraints, a maximum of six service staff can be employed at any one time.
 - Experimental factors:
 - Staff roster
 - Responses:
 - % of customers queuing for less than 3 minutes
 - Histogram of waiting time for each customer in the queue
 - Time series of mean queue size by hour

• Staff utilisation Nottingham

UNITED KINGDOM · CHINA · MALAYSIA

- 4. Design the conceptual model: The model content
 - Model must be able to accept the experimental factors and to provide the required responses
 - Scope of the model must be sufficient to provide link between the experimental factors and responses
 - Scope of the model must also include any other processes that have a significant impact on the response
 - Level of detail must be such that it represents the components defined within the scope and their interconnections with sufficient accuracy



- 4. Design the conceptual model: The model content (cont.)
 - Use rapid prototyping throw away models to decide about scope and level of detail
- Keep a record of all assumptions that are made during the design of the model content!!!





ASSS 2014



• Case Study: Fast-Food Restaurant

| Model Scope | Detail | Decision | Justification |
|--------------------------|------------------|----------|---------------|
| Customers | | | |
| Staff | Service | | |
| | Food preparation | | |
| | Cleaning | | |
| Queue at service counter | | | |
| Tables | | | |
| Kitchen | | | |

| Model Level of Detail | Detail | Decision | Comments (Details) |
|-----------------------|-----------------------|----------|--------------------|
| Customers | Inter-arrival time | | |
| | Size of order | | |
| Service staff | Service time | | |
| | Staff rosters | | |
| | Absenteeism | | |
| Queues | Queuing | | |
| | Capacity | | |
| | Queue behaviour | - | - |
| | - jockey, balk, leave | | |
| | - join shortest queue | | |





• Case Study: Fast-Food Restaurant

| Model Scope | Detail | Decision | Justification |
|--------------------------|------------------|----------|-----------------------------------|
| Customers | | Include | Flow through service process |
| Staff | Service | Include | Required for response |
| | Food preparation | Exclude | Material shortage not significant |
| | Cleaning | Exclude | Not related to speed of service |
| Queue at service counter | | Include | Required for response |
| Tables | | Exclude | Not related to customers waiting |
| Kitchen | | Exclude | Material shortage not significant |

| Model Level of Detail | Detail | Decision | Comments (Details) |
|-----------------------|-----------------------|----------|---------------------------------------|
| Customers | Inter-arrival time | Include | Distribution |
| | Size of order | Exclude | Represented in service time |
| Service staff | Service time | Include | Distribution |
| | Staff rosters | Include | Experimental factor |
| | Absenteeism | Exclude | Could be represented in staff rosters |
| Queues | Queuing | Include | Required for responses |
| | Capacity | Exclude | Assumption: unlimited |
| | Queue behaviour | - | - |
| | - jockey, balk, leave | Exclude | Not well understood |
| | - join shortest queue | Include | Well understood |



Graphical Representation





The role of data in conceptual modelling



- Data for model realisation are not required for conceptual modelling, but are identified by the conceptual model
- Sometimes it is difficult or even impossible to obtain adequate data making the proposed conceptual model problematic!
- What can you do in these cases?
 - Redesign the conceptual model and leave out the troublesome data
 - Estimate the data
 - Treat data as an experimental factor rather than a fixed parameter



Summary



• What did you learn?



Further Reading

- Further Reading
 - Robinson (2008a; 2008b)
 - Bommel and Müller (2008)
 - Robinson et al. (2010)



- Acknowledgement
 - The content of this presentation is a summary of Robinson (2004) chapter 5 and 6



Questions / Comments





References

- Bommel P and Müller JP (2008). An Introduction to UML for Modelling in the Human and Social Sciences. In: Phan D and Amblard F (Eds.) Agent-based Modelling and Simulation in the Social and Human Sciences.
- Jensen K, Kristensen L M, and Wells L (2007). Coloured petri nets and CPN tools for modelling and validation of concurrent systems. International Journal on Software Tools for Technology Transfer, 9(3):213-254
- Pels H J and Goossenaerts J (2007). A Conceptual Modeling Technique for Discrete Event Simulation of Operational Processes. IFIP International Federation for Information Processing, 246 pp. 305-312
- Robinson S (2004). Simulation: The practice of model development and use.
- Robinson S (2008a). Conceptual modelling for simulation Part I: Definition and requirements, JORS, 59(3):278-290
- Robinson S (2008b). Conceptual modelling for simulation Part II: A framework for conceptual modelling, JORS, 59(3):291-304
- Robinson S, Brooks R, Kotiadis K, and Van Der Zee D-J (Eds.) (2010). Conceptual Modeling for Discrete-Event Simulation.

