

An aerial photograph of the TU/e campus at dusk. The main building, a tall glass skyscraper, is brightly lit from within, with its lights reflecting on the surrounding trees and other buildings. The sky is a mix of orange and blue. In the foreground, there are more buildings and a road, all partially obscured by a red semi-transparent overlay.

Business Process Simulation

Lecture 1a - Introduction to simulation

Laura Genga

Overview on lecture modules

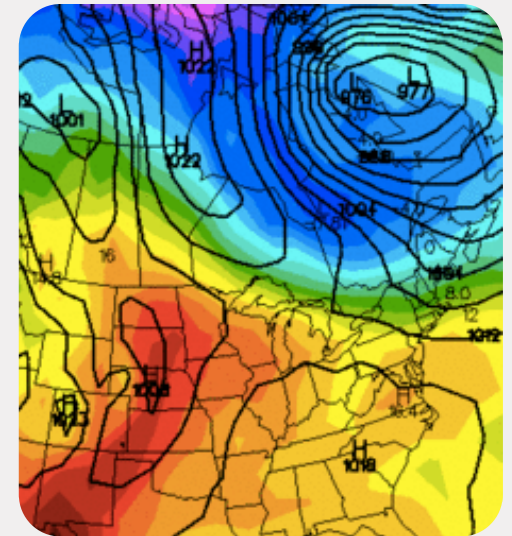
- a) Introduction to simulation
- b) Step 1: Project definition
- c) Step 2: Design the study

Simulation

*Simulation is the **imitation** of some real thing, state of affairs, or process. The act of simulating something generally entails representing certain **key characteristics or behaviours** of a selected physical or abstract system*

Simulation

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Definitions of simulation

The art and science of **creating a representation** of a process or system for the purpose of **experimentation** and evaluation

(Gogg and Mott, 1996)

Computer simulation is the process of designing a mathematical-logical **model of a real system** and **experimenting** with this model on a computer

(Pritsker, 1986)

Simulation is the **imitation** of the operation of a real-world process or system over time

(Banks, Carson, & Nelson, 1996)

Simulation refers to a broad collection of methods and applications to **mimic** the behavior of real systems, usually on a **computer** with appropriate software

(Kelton, Sadowski & Sadowski, 1998)

Model development

Creating an abstraction (model) of the real world.

Enabling us to draw conclusions about the real system by studying and analyzing the model.

The physical model

Emergency 'drills'

Dress rehearsal

Test runs of a new attraction



The scaled model

Small scale wind tunnel models

Physical flight simulators to train pilots

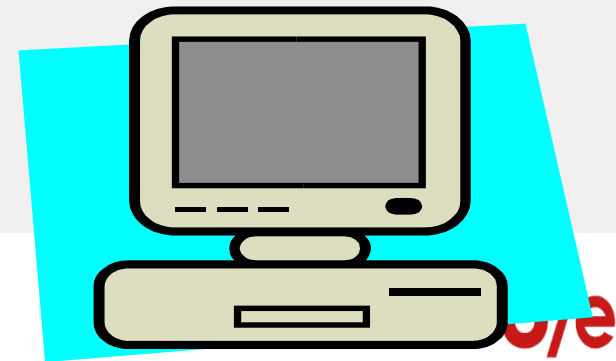
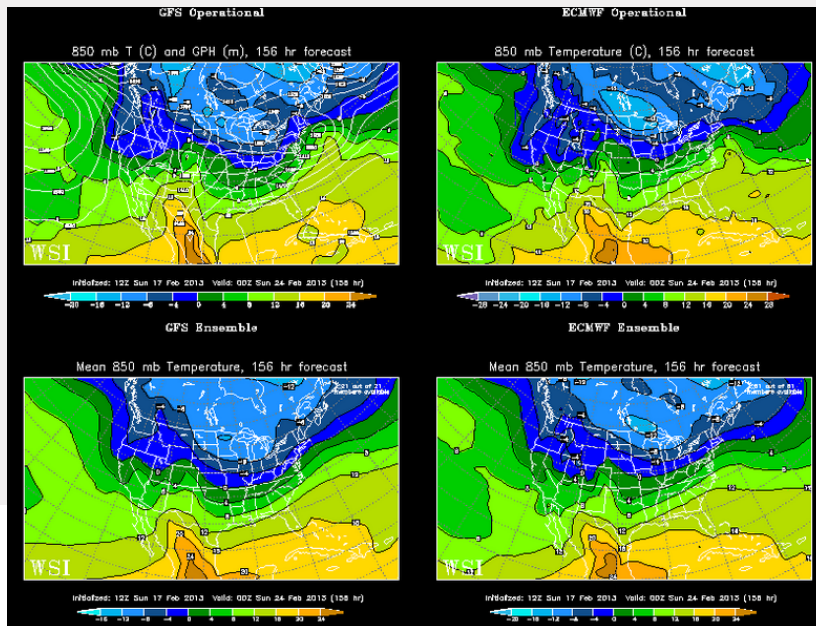
Tabletop models of material handling systems

Model trains and railways

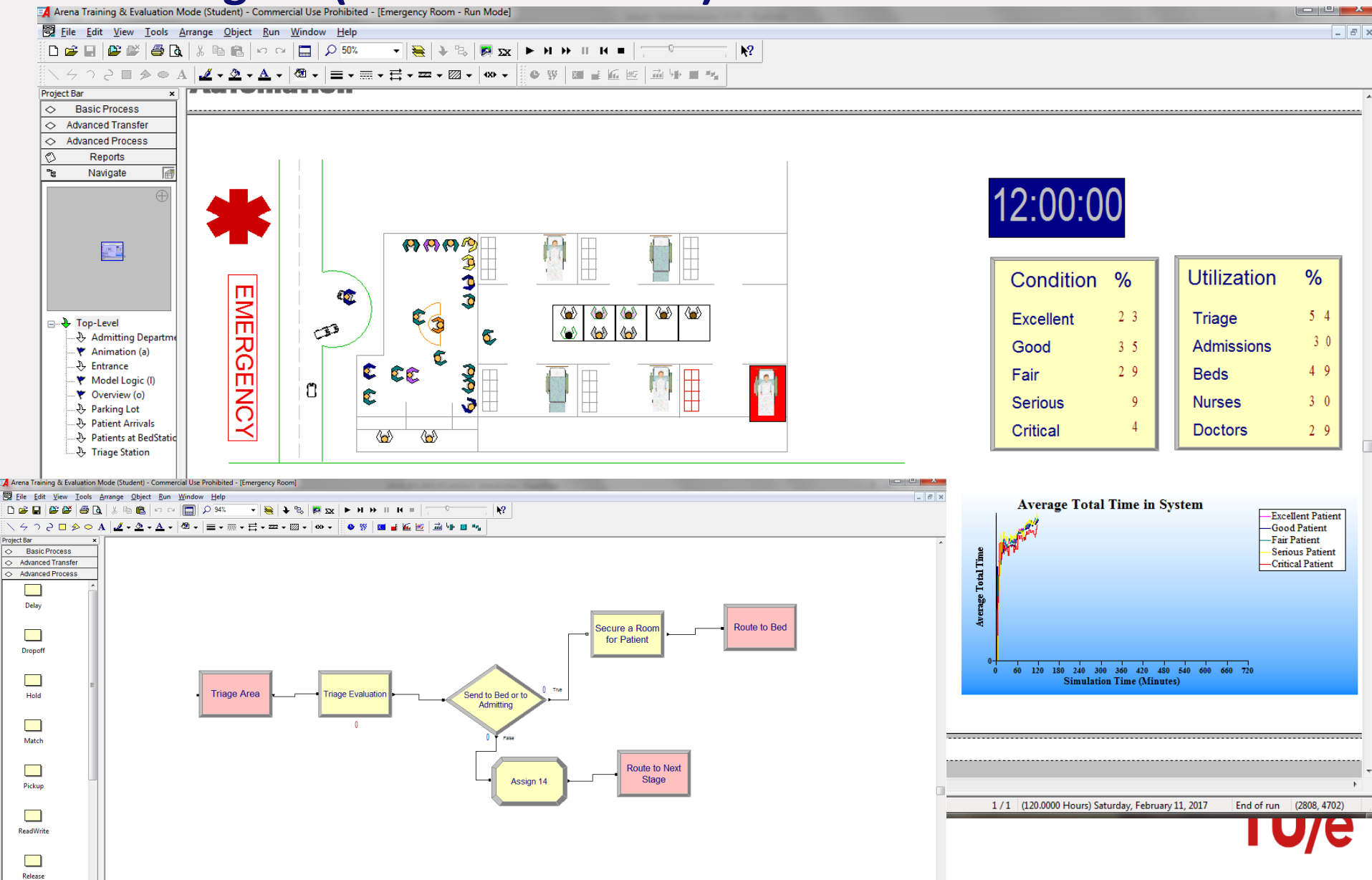


The logical (or mathematical) model

mathematical approximations of the system
usually represented in a computer program
easy, cheap, and fast



The logical (or mathematical) model



Why do people use simulation models?

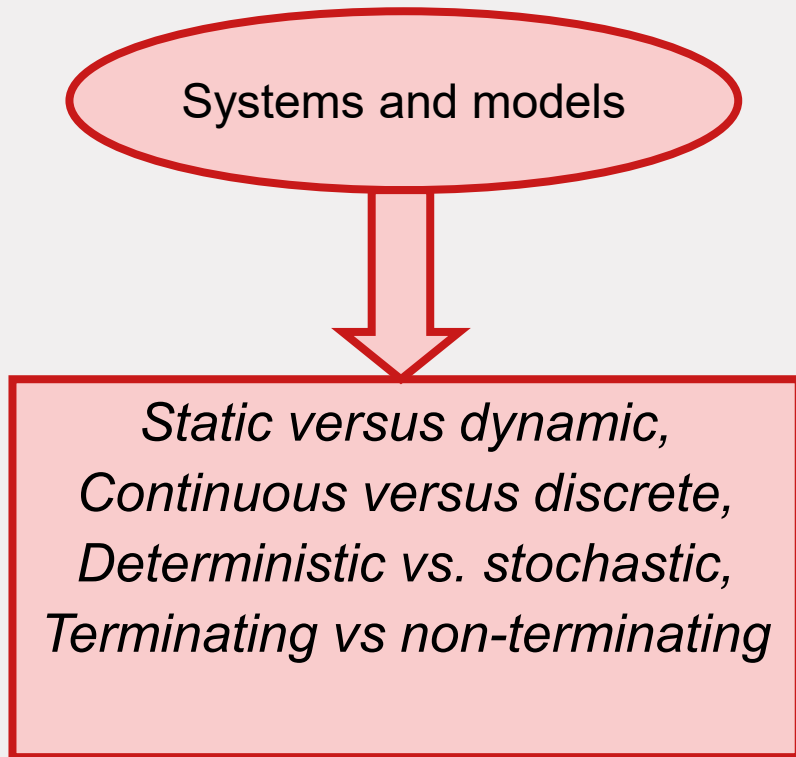
Most systems are stochastic and dynamic

Analytical model not feasible and/or not practical.

Trial by implementation can be too costly, or too time consuming, or even too dangerous.

Timescale may be too long to get accurate real measurements.

Different kinds of systems



Static versus dynamic

A static model

- Not dependent on time
 - A dice

A dynamic model

- Time dependent
 - Service and manufacturing related systems
 - A few examples of dynamic variables:
 - production schedules,
 - equipments utilizations,
 - customer arrival rates and processing times

Discrete-event versus Continuous

Discrete-event simulation

- the state variables change instantaneously at separate points in time.
- time is advanced in discrete “jumps” as the simulation advances from event to event
- e.g. orders handled by servers.

Continuous simulation

- the state variables change continuously with respect to time.
- time is advanced in (almost) infinitely small increments
- e.g. mass-spring system

Discrete-event versus Continuous

System	Discrete events	Continuous events
The flow of customers through a bank's drive-in teller	X	
The consumption of fuel by a fighter jet during routine flight manoeuvres		X
The spread of cancer through a victim's lymphatic system		X
The assembly of cars at an automobile assembly plant	X	

Stochastic versus deterministic

A stochastic model contains processes controlled by random variables.

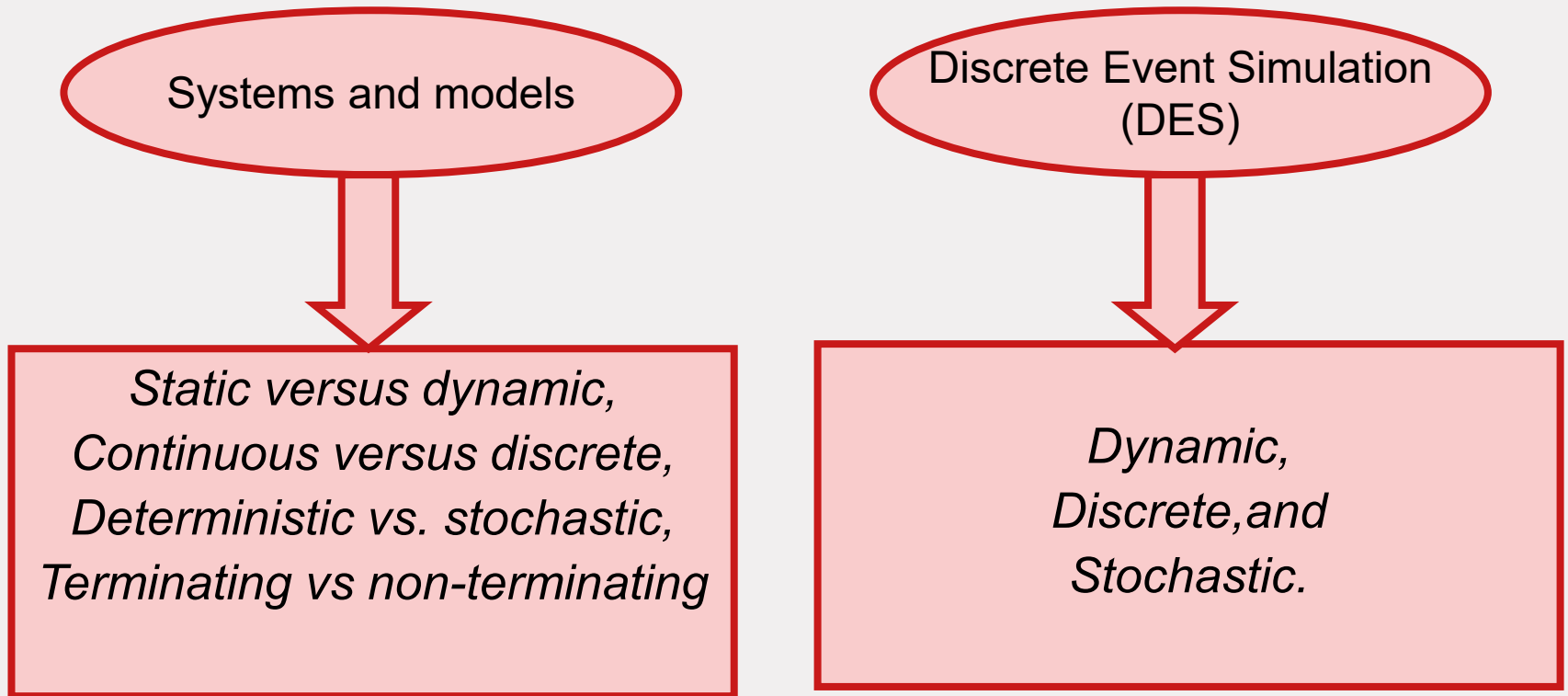
The word *variables* implies that something is capable of changing. It does not have a specific value, but rather a range of values.

Random signifies that the changes can occur with no particular pattern.

A stochastic process is composed of a sequence of randomly determined values.

A deterministic model does not contain random variables and is not influenced by probability

Different kinds of systems



An aerial night photograph of the TU/e campus in Eindhoven, featuring several modern glass-fronted buildings with interior lights glowing. The image is partially covered by a semi-transparent red rectangle that serves as a background for the text.

Business Process Simulation

Lecture 1b - Step 1: Project definition

Laura Genga

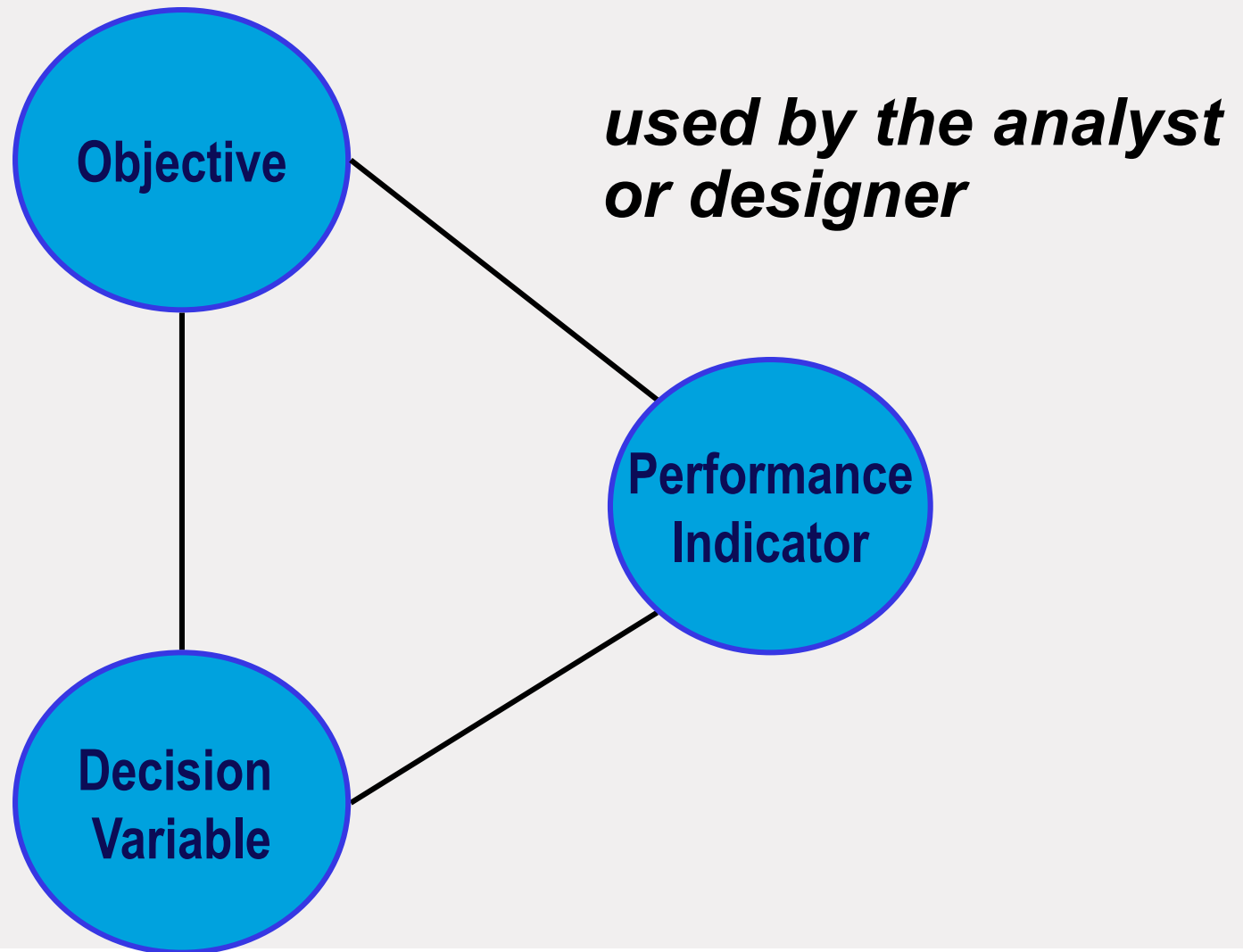
Overview on lecture modules

- a) Introduction to simulation
- b) Step 1: Project definition**
- c) Step 2: Design the study

STEP 1: Project definition

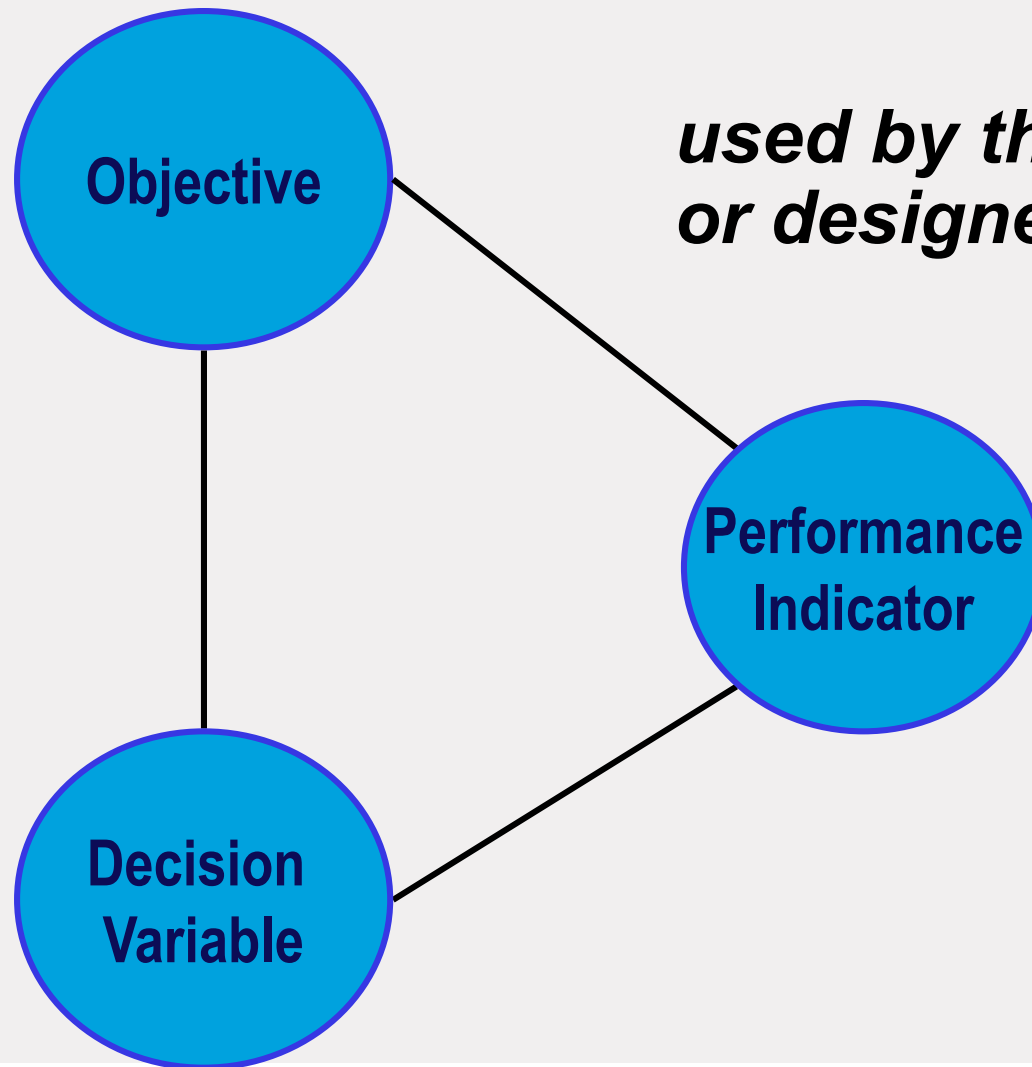
1. Problem formulation
2. Phrase research questions
3. Set the scope
4. (Estimate required resources needed to do study)
5. (Perform a cost-benefit analysis)
6. (Create a planning of the proposed project)

Decision Framework

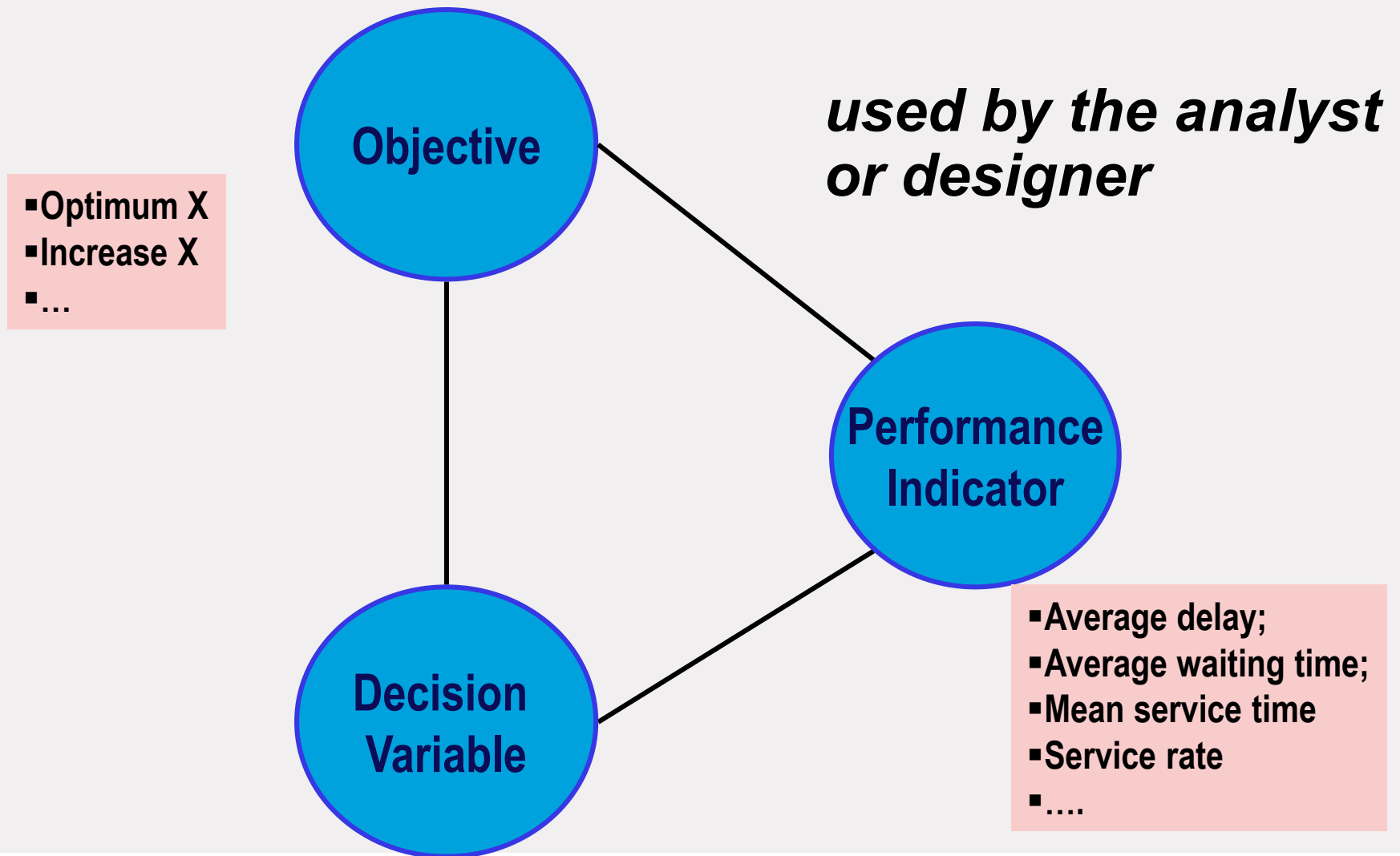


Decision Framework

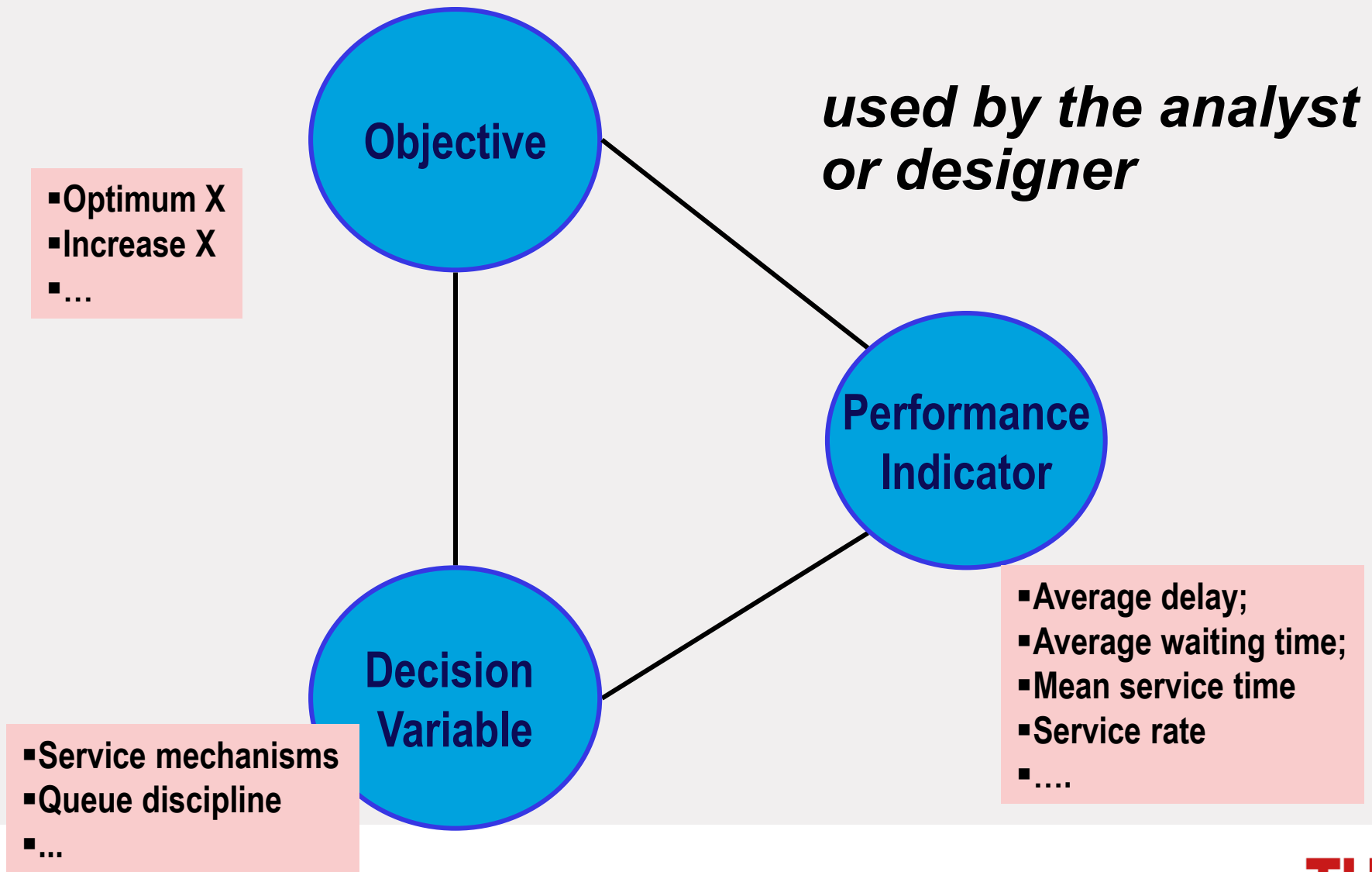
- Optimum X
- Increase X
- ...

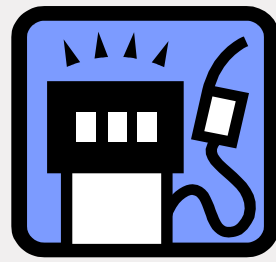


Decision Framework

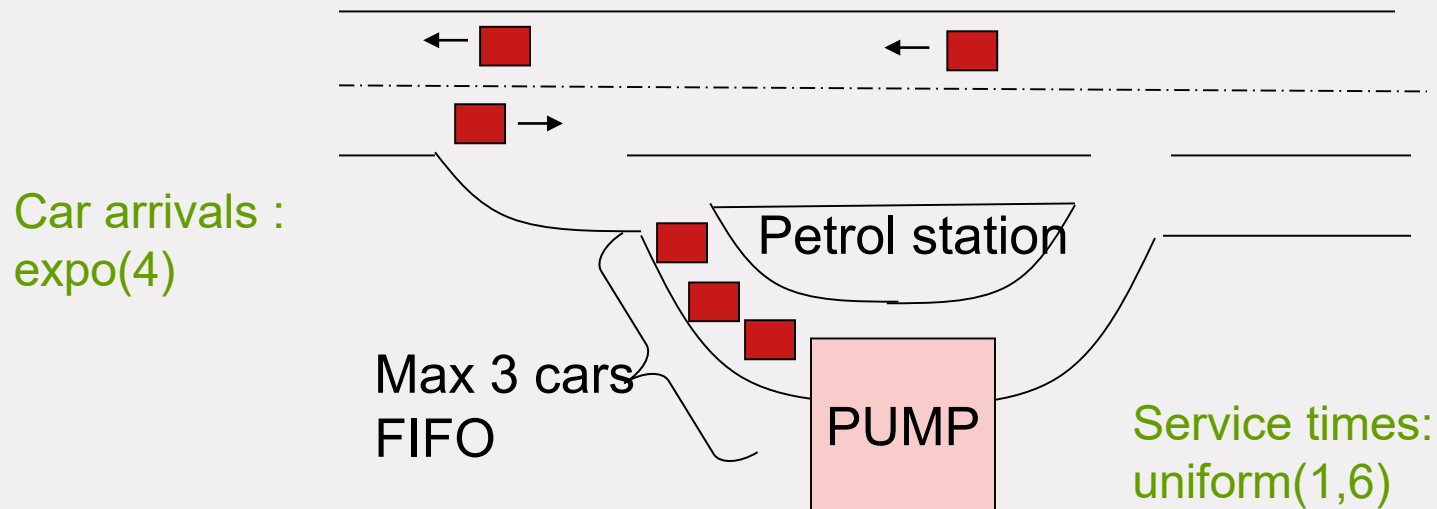


Decision Framework

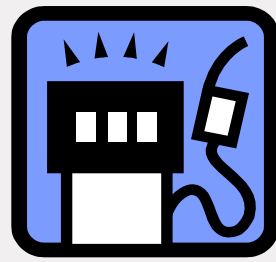




EXAMPLE: The Petrol Station

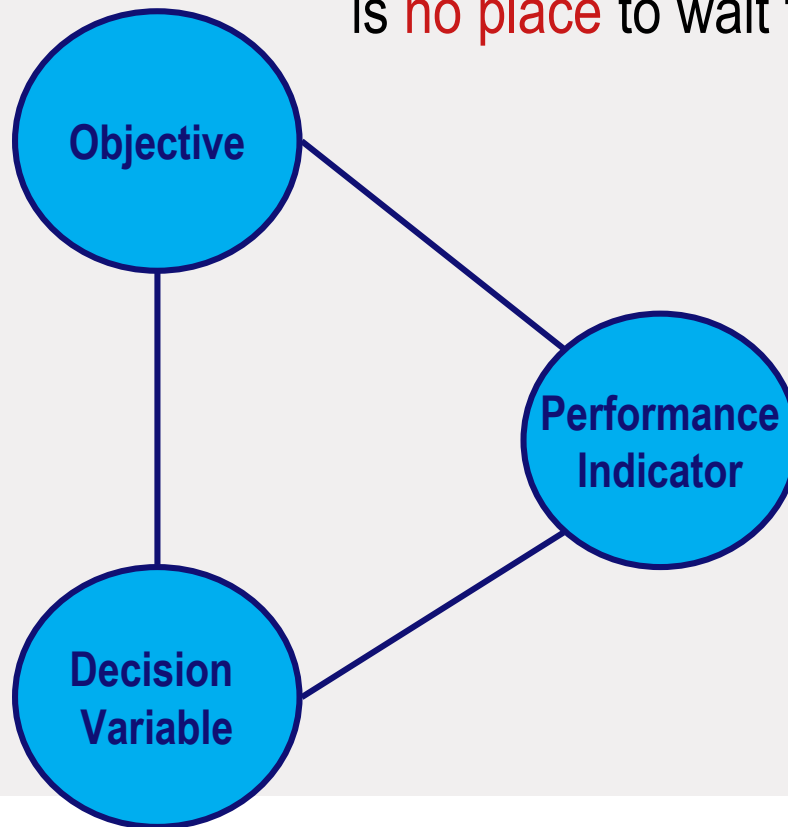


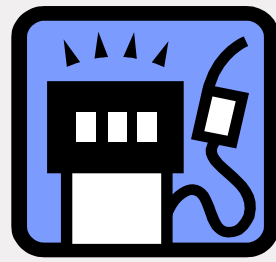
The owner of the petrol station has the feeling that some potential customers are leaving the station because there is **no place** to wait for service



Step 1.1: Decision Frame

...the feeling that some potential customers are leaving the station because there is **no place** to wait for service

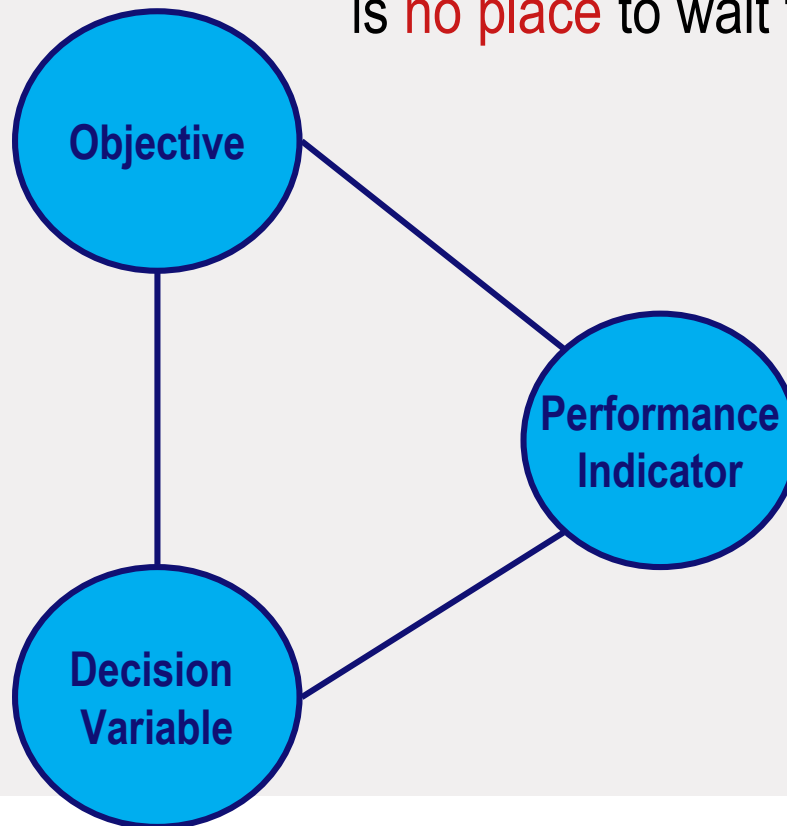


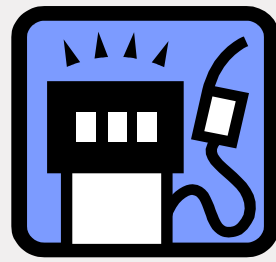


Step 1.1: Decision Frame

- Less customers driving on

...the feeling that some potential customers are leaving the station because there is **no place** to wait for service

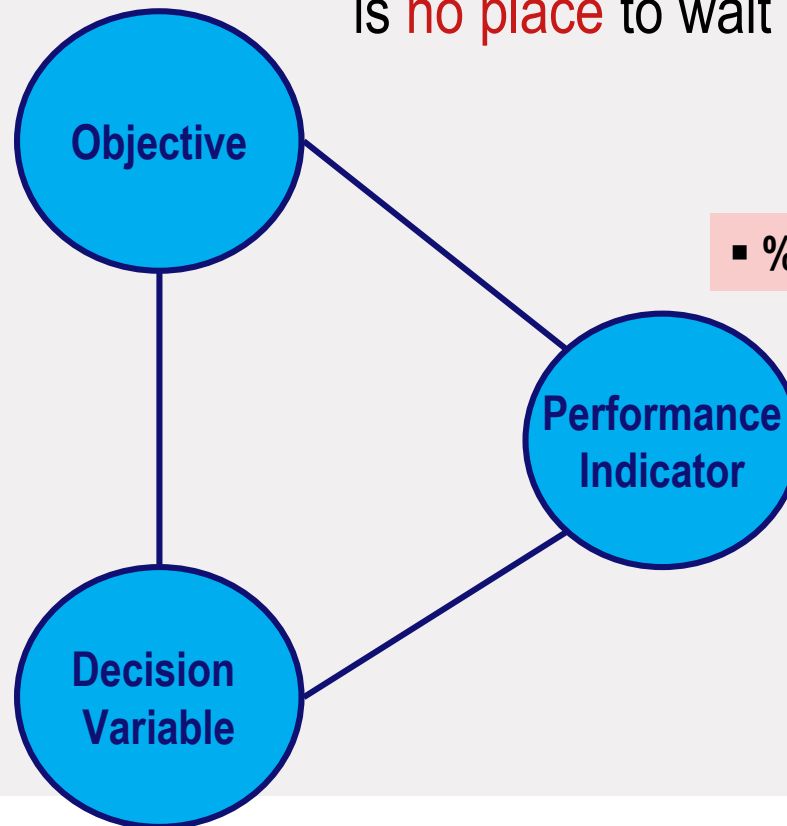




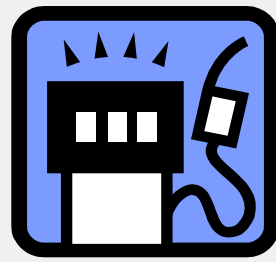
Step 1.1: Decision Frame

- Less customers driving on

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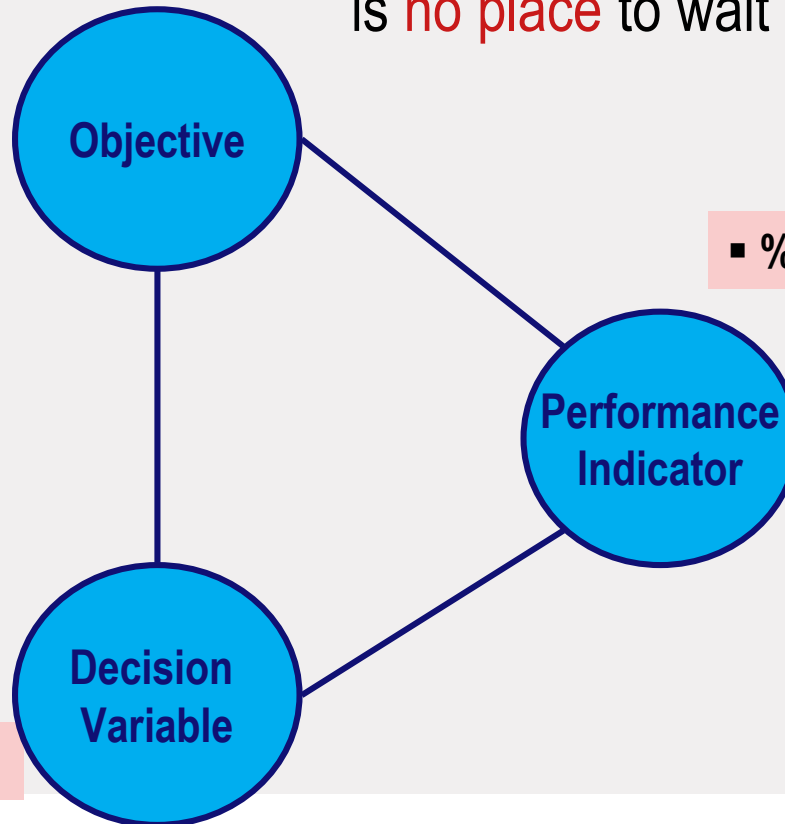
- % of customers driving on



Step 1.1: Decision Frame

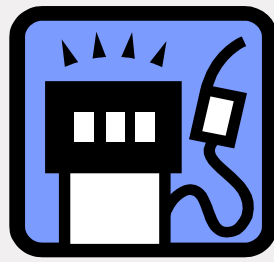
...the feeling that some potential customers are leaving the station because there is **no place** to wait for service

- Less customers driving on



- % of customers driving on

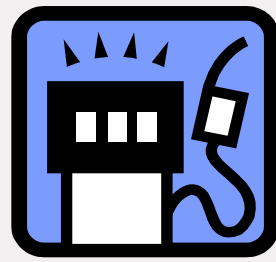
- size of queue area



Step 1.2: Research Questions

What is the percentage of customers that drive on without being served?

What is the effect of increasing the capacity of the queue area on the percentage of cars leaving without being served?



Step 1.3 Scope

Included in the model:

- The pump with service start and service end events.
- The queue area with arrival, departure events
- The cars with the events matching the queue area and petrol events.
- The decision of the car to drive on when it finds the queue area full.

Excluded from the model:

- The time for paying
- The volume of gas tanked
- The type of car
- The shop
- etc.

Manager's perspective on decision

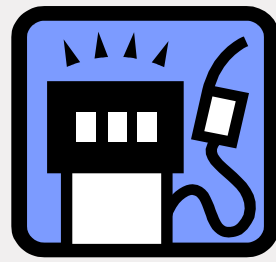
Company loses money when customer leaves

- Each customer that leaves costs ?

Does an investment in more queueing places pay off?

- How much does the additional queueing place cost?
- When should this be earned back? After one year?
- Doesn't the waiting time increase too much?

We can only measure number of customers that leave ...



Alternative decision frame 2

Suppose:

- One extra queueing place costs 750 euros
- Each customer leaving 'costs' 100 euros
- Manager wants return on investment after 1000 customers

Decision Frame:

- **Objective:** earn more money by serving more customers (decrease the number of customers leaving because of full queue by extending the queue area)
- **Decision variables:** capacity of the queue area
- **Performance measures:** money lost by customers driving on because of a full queue:
 - number of customers lost * 100 euro's +
 - number of additional queueing places * 750 euro's

Exercise: the harbour case

A harbour can host two types of ships; Small and Big ships, which arrive with an interarrival time exponentially distributed with a mean of 5,5 hours and 6.7 hours respectively. There are two docks (dock1 and dock2) where ships can be unloaded.

Small ships (big ships) are unloaded at dock1 (dock2) with a service time uniformly distributed between 3 and 7 (2 and 8).

If dock1 is empty and there are Big ships waiting at dock2 then a Big ship can go to dock1 and is served with $1,5 * \text{Uniform}(2,8)$. If dock2 is empty and there are Small ships waiting at dock1 then a Small ship can go to dock2 and is served with $2 * \text{Uniform}(3,7)$. For both docks the queue discipline is SPT (Shortest processing time first).

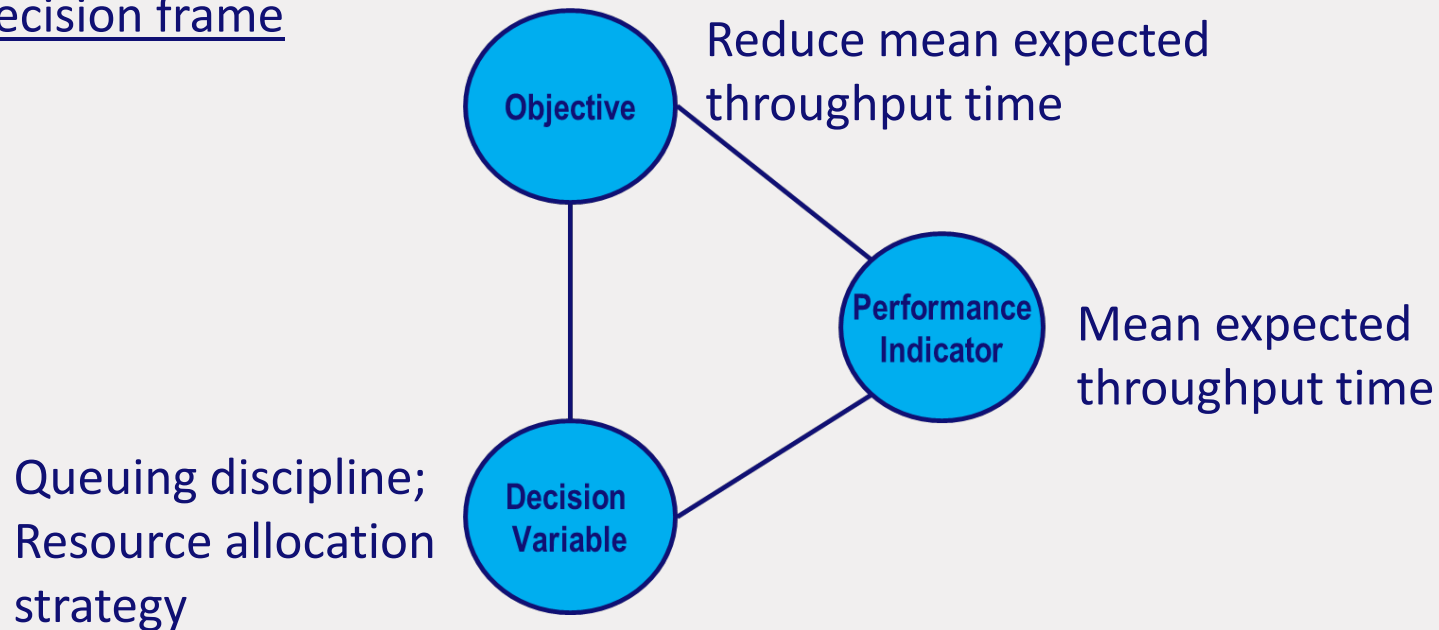
The management team of the harbour would like to improve the mean expected throughput time. Possible options are closing dock1 to Big ships and dock2 to small Ships, or to change the queueing discipline.

Formulate a decision frame for this problem, together with its research questions and scope

Proposed solution

Problem formulation. Waiting by ships is expensive for their owner-operators. For this reasons harbours could explore options to reduce waiting time.

Decision frame



Proposed solution (continue)

Research questions

1. Does closing Dock1 to Big Ships and Dock2 to Small Ships improve the mean expected throughput time of ships at the harbour?
2. Is it better to use FIFO rule instead of SPT rule?

Scope

- The docks with service start and service end events.
- The queue with arrival, change queue and service start events.
- The ships with the events matching the queue and dock events.

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Business Process Simulation

Lecture 1c - Step 2: Design the study

Laura Genga

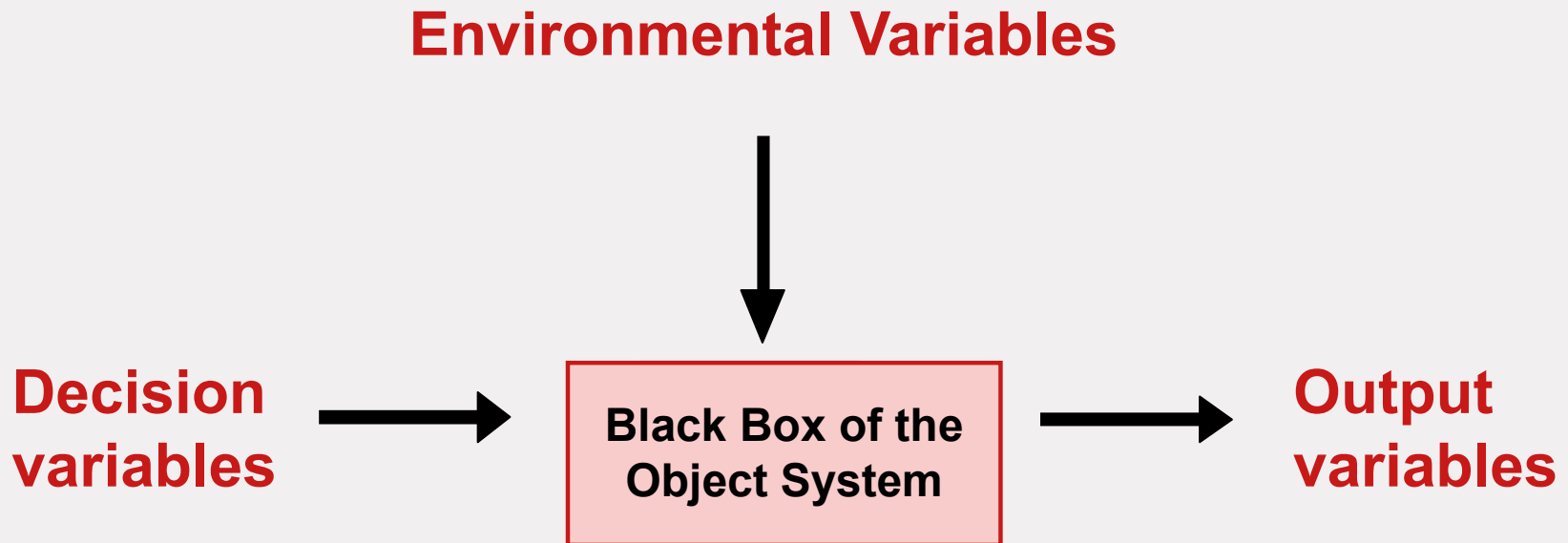
Overview on lecture modules

- a) Introduction to simulation
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STEP 2: Design the Study

1. Create **black box** representation.
2. List **assumptions**
3. Determine if a simulation model is actually needed
4. Estimate the **number of models** required.
5. Make a detailed planning
6. (Select the tool: done, it is ARENA)
7. (Collect data: it is given; identify the relevant data)

Step 2.1: black box representation



In 't Veld (1CK10):

A black box representation is a representation of a system or sub system of which the internal components and relationships are not (yet) known to the analyst.

Decision Variables

Other names:

- decision variables (ISO 15704)
- control variables (Bertrand/Fransoo)
- design/control variables (Kulkarni)
- 'stuurvariabelen' (Griep/Flapper):
- 'beslissingsvariabelen' (Griep/Flapper)
- design parameters (Cochrane)

Used to consciously influence the system

Examples:

- Queue discipline: FIFO or SPT?
- Resource allocation strategies

Output Variables

Other names:

- performance variables (Bertrand, Fransoo)
- performance indicators (ISO 15704)
- key performance indicators

Used to measure the performance of the system

Examples:

- Mean waiting time
- Service time of a car

Law & Kelton Chapter 9: Output Analysis

Environmental Variables

Other names:

- Input variables
- 'omgevingsvariabelen' (Griep/Flapper):

Influence the system (are givens)

Examples:

- Car interarrival time
- Service time of a car

Law & Kelton Chapter 6: Input Analysis

Black Box Representation

Environmental Variables

e.g.:

- Interarrival time of clients
- Service time
-



Decision variables

e.g.:

- Service mechanism
- Queue discipline
-

**Black Box of the
Object System**

Output variables

e.g.:

- Mean waiting time
- % not served
-

Step 2.2: Assumptions and Givens

A formal list must be maintained and approved by the stakeholders (possibly validated)

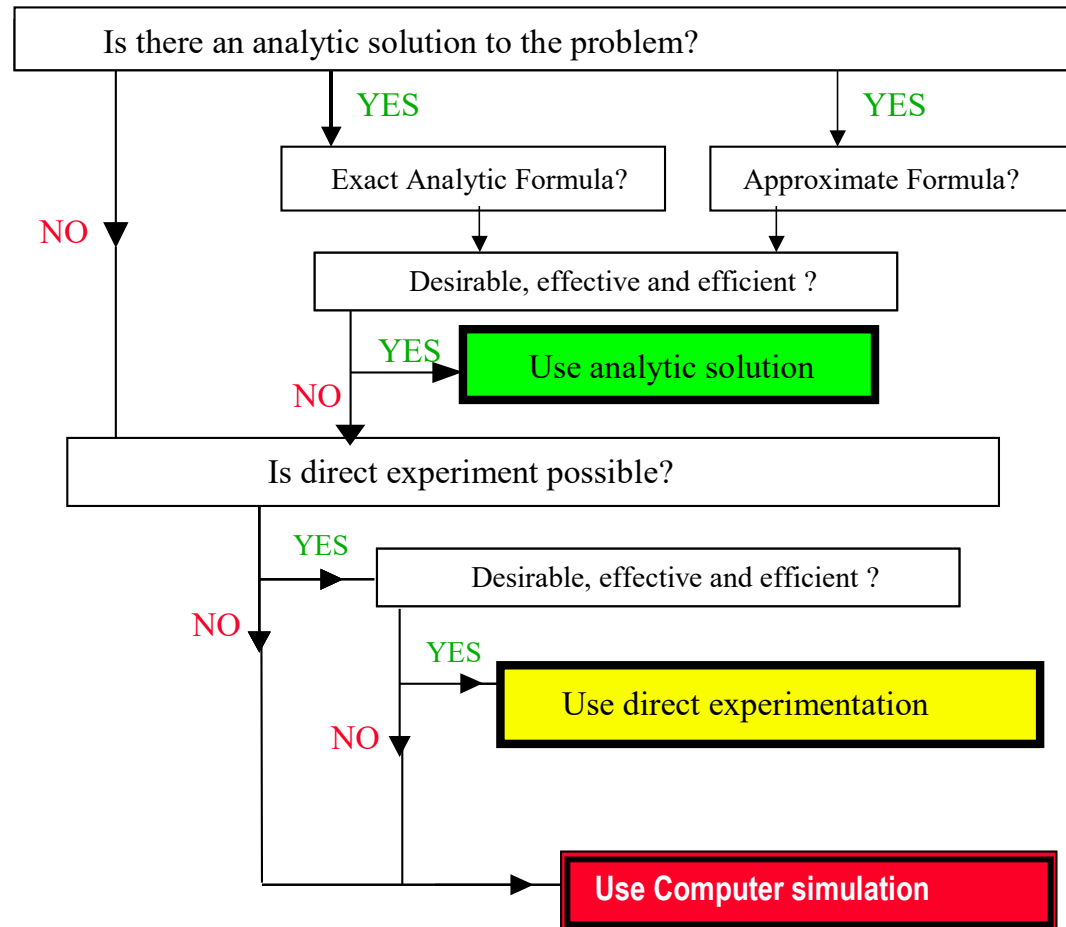
Assumptions determine:

- The scope of the work
- Micro-level decisions in models depend on them

TIP:

Keep a list of all assumptions you make throughout the steps of your project !

Step 2.3: Is simulation suitable?

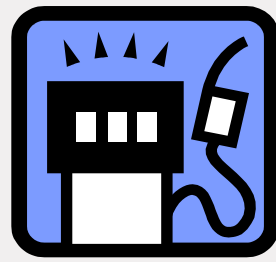


Step 2.4 Number of models

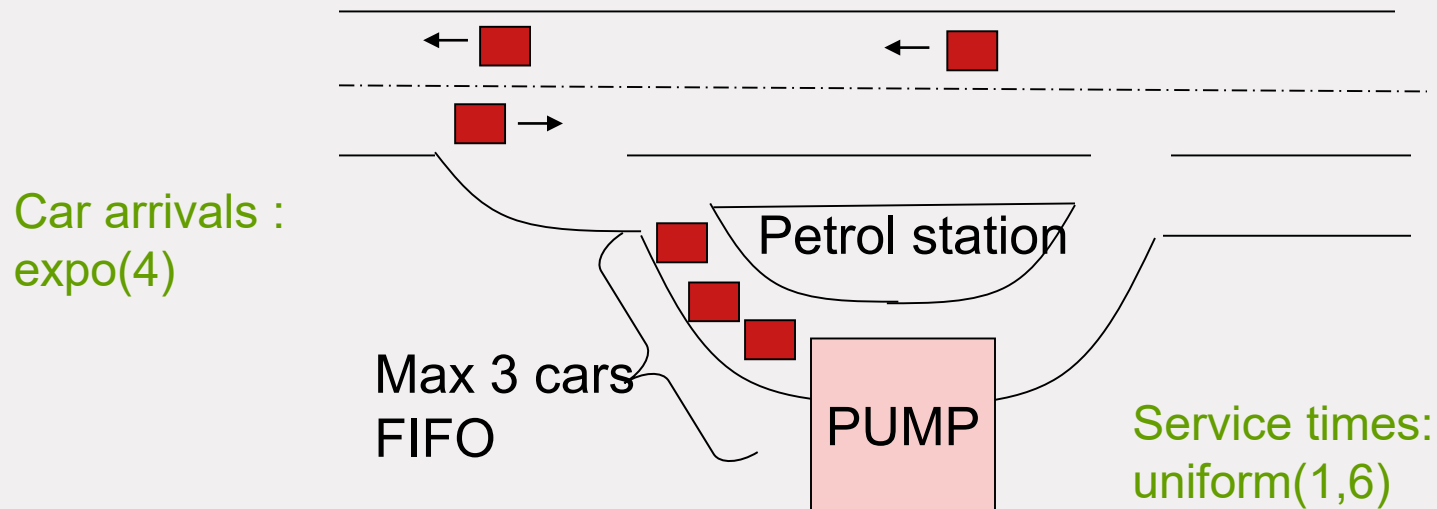
Different variants / designs?

Different parameter settings?

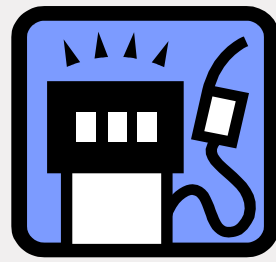
Simulation / validation?



EXAMPLE: The Petrol Station



The owner of the petrol station has the feeling that some potential clients are leaving the station because there is **no place** to wait for service



Step 2.1: Black Box

Environmental Variables

- interarrival time of cars
- service time of cars



Decision variables

- Size of the queue area

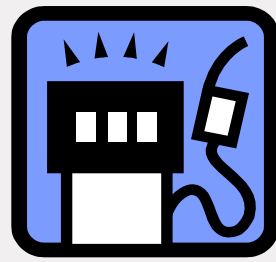


PETROL STATION

Output variables

- percentage not served
- (mean waiting time)
- (number not served)





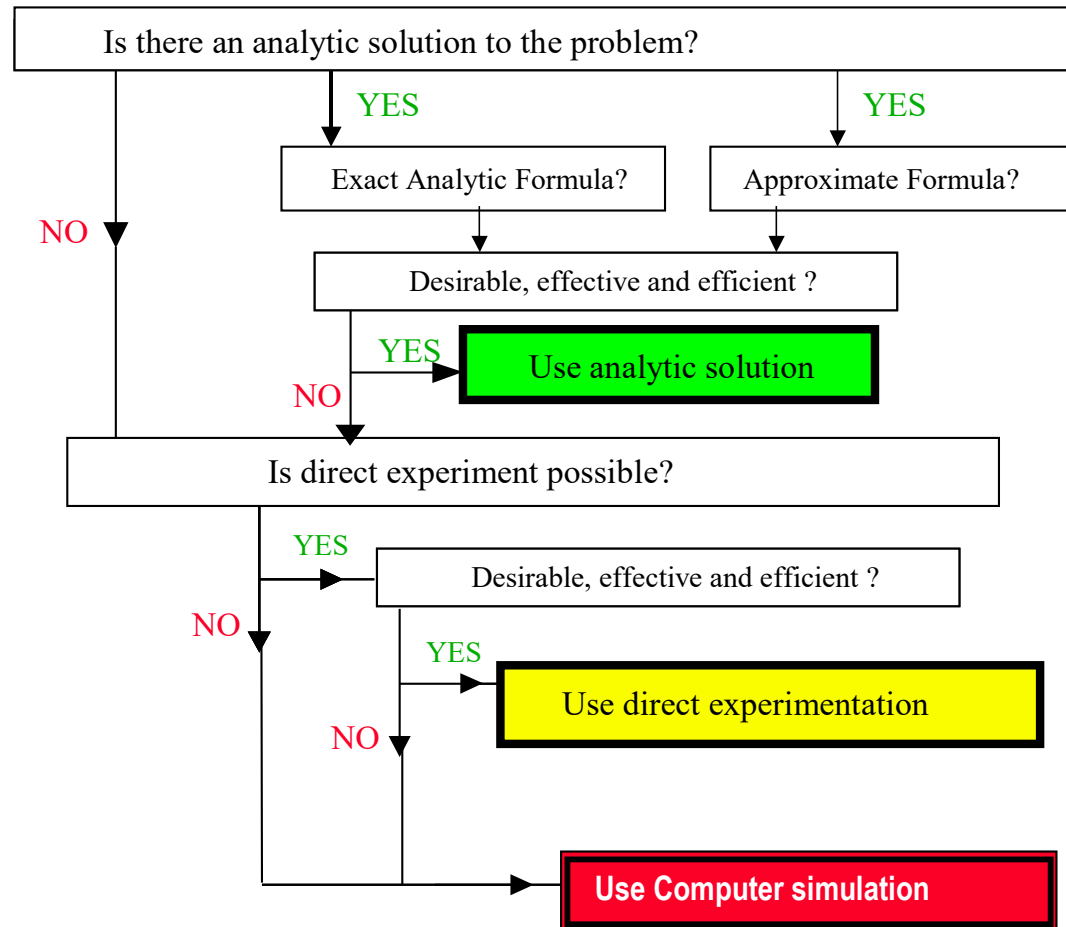
Step 2.2: Assumptions and givens

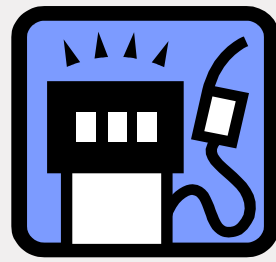
- G1. the interarrival time of cars is $\text{expo}(4)$
- G2. the service time of cars is $\text{uniform}(1,6)$
- G3. queue capacity is 3 cars (plus 1 car in service)
- G4. queueing discipline is FIFO

- A1. the business operates 24 hours per day; 7 days per week
- A2. no defects of the pumps occur
- A3. time of payment can be incorporated into processing time

List of assumptions must be maintained as the study (and the modelling) proceeds

Step 2.3: Is simulation suitable?





Step 2.3: Simulation suitable?

Characteristics of the system:

- Single class system
- No admission control
- One queue (FIFO) with limited capacity ($3+1 = 4$)
- One server
- A/B/M/K/N notation? \Rightarrow M/G/1/4 system

Are there analytical methods available for M/G/1/4?

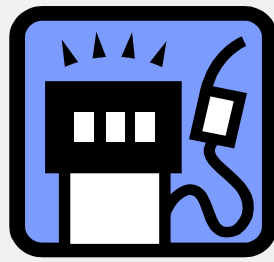
- No!

Can we make any approximation?

- Yes: M/M/1/4 system, but not precise enough!

Is direct experimentation possible?

- No!



Step 2.4: Number of models

For the problem owner:

1. current situation (queue length = 3)
2. situation with longer queue
 - a) Queue length = 4
 - b) Queue length = 5
 - c) Queue length = 6
 - d) ... ?

Exercise: the harbour case

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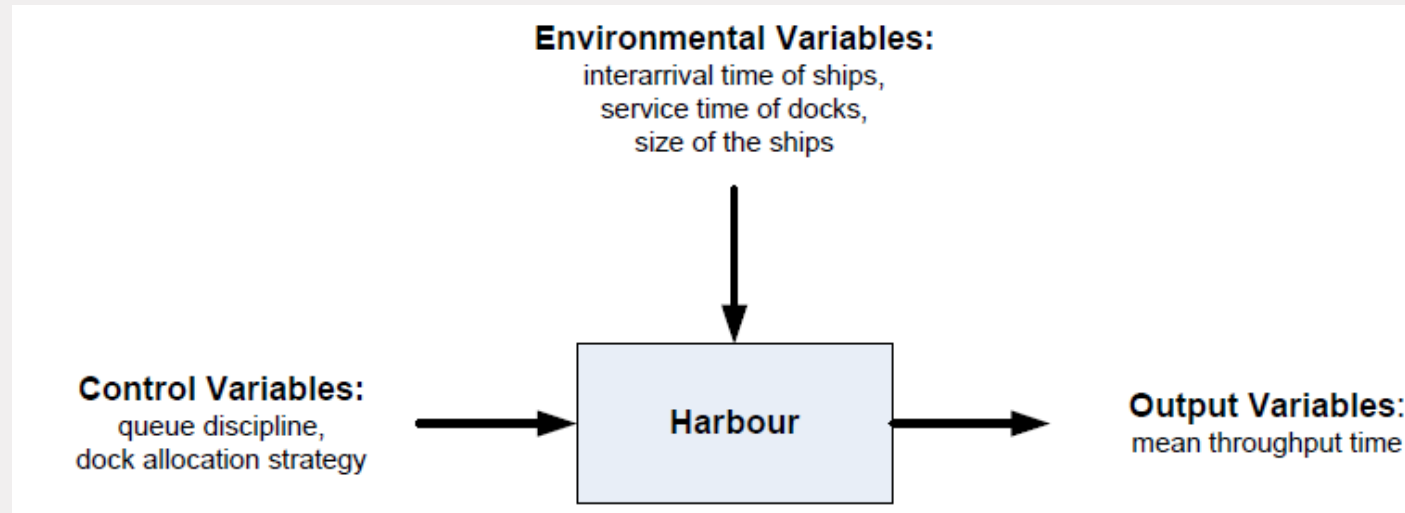
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The management team of the harbour would like to improve the mean expected throughput time. Possible options are closing dock1 to Big ships and dock2 to small Ships, or to change the queueing discipline.

Draw a black box representation for this problem. Is simulation needed? Why?

Proposed solution



Is simulation needed?

Yes. Changing queues cannot be addressed in analytic models