A DISCRETE-EVENT SYSTEM APPROACH TO MULTI-AGENT DISTRIBUTED CONTROL OF CONTAINER TERMINALS

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Keywords: Multi-Agent Systems, Discrete Event Systems, Container Terminals, Transport Systems.

Abstract: The area of modelling and controlling intermodal terminal systems is relatively new. The paradigms of

Discrete Event Systems for modelling and of Multi-Agent Systems (MAS) for distributed control seem promising. Many research attempts developed modelling and simulation tools but no standard exists. This paper presents a Discrete Event System model of the agents introduced to describe how a distributed control of the terminal activities can be achieved. The interactions between four classes of agents are detailed. The approach is useful to develop a simulation platform to test MAS efficiency in terminal management and to

measure the performance of static or adapted control strategies.

1 INTRODUCTION

Planning and scheduling resources in a maritime container terminal pose very complex modelling and control problems to the scientific community. Flexible and powerful modelling and simulation tools are necessary to describe an intermodal hub system made up of many different infrastructures and services. In the absence of standard tools, several research studies are based on discrete event simulation techniques (Vis and De Koster, 2003), on mathematical models or empirical studies (Crainic *et al.*, 1993, Gambardella *et al.*, 1998).

In the context of intelligent control of transport systems, this paper uses the Discrete EVent System (DEVS) specification technique (Zeigler et al., 2000) to completely and unambiguously characterize a Multi-Agent System (MAS) for controlling a container terminal. Autonomous agents play as atomic dynamic DEVS. They exchange messages one with another to negotiate services in a common environment. The DEVS approach leads to heterarchical MAS where all information and control functions are distributed across agents, and allows rigorous theoretical analysis of structural properties of the MAS. Moreover, the DEVS formalism is an interesting alternative to other tools for MAS specification (Huhns and Stephens, 2001, Lin and Norrie, 2001). It is suitable to develop models both for simulation and for implementation of the software controllers.

As in MAS for manufacturing control (Heragu, 2002), agents can use decision algorithms based on a fictitious currency to buy services from other seller agents which use pricing strategies. Sellers and buyers reach an equilibrium between conflicting objectives, i.e. to maximize profit and to minimize costs, respectively. Recent analytical models of negotiation (Hsieh, 2004) underline the need of a systematical analysis and validation method for distributed networks of autonomous agents. Other researches focus on the experimental and detailed validation of MAS on distributed simulation platforms (Shattenberg and Uhrmacher, 2001, Logan and Theodoropoulos, 2001).

Here, a detailed DEVS model is developed for the interactions between the agents concurrently operating for the critical downloading process of containers from a ship. This paper is a contribution to define a complete DEVS model to develop a simulation platform for testing and comparing the proposed MAS with other distributed control architectures for container terminals. The simulation model could be used to design and test alternative system layouts and different control policies, both in standard and perturbed operating conditions.

The paper is organized as follows. Section 2 describes the main processes in a terminal. Then, the basic components of the MAS are presented, and their roles and relations are indicated. Section 3 models agents as atomic DEVS, and focuses on their interactions when containers are downloaded from

ships to the terminal yard. Section 4 gives some ideas about the simulation platform to test efficiency and robustness of the proposed MAS architecture. Section 5 overviews the benefits of the approach and enlightens open issues.

2 THE MAS FRAMEWORK FOR CONTAINER TERMINALS

To define the autonomous agents operating and interacting in a terminal environment, the main terminal processes have to be abstracted. They are commonly identified as import, export, and transshipment cycles.

2.1 Import, Export and Transshipment Cycles in a Container Terminal

Imported containers arrive on a vessel or feeder ship and depart by trains/trucks. Exported containers arrive by trains/trucks and depart on a vessel ship (transition from railway/road to sea mode). Transshipped containers arrive and depart by ship: cargo is moved from vessel (feeder) to feeder (vessel) ships for close (far) destinations.

Then, containers have to be: loaded/downloaded on/from ships; delivered/picked to/from trucks or trains; stacked and kept in blocks, in which the terminal yard is divided; transferred from ship/train/truck to blocks and backwards; consolidated, i.e. redistributed between blocks to allow fast retrieval. To this aim, several dedicated or shared resources are used: quay cranes, yard cranes, railway cranes; internal transport vehicles (trailers); quays and berths, lanes for internal transport, gates, railway tracks; skilled human operators.

A road import cycle follows three steps. Firstly, quay cranes download containers from a berthed ship to trailers, which transfer cargo to yard. Here, yard cranes pick-up containers from trailers and stack them in assigned positions. Secondly, containers stay in the blocks while waiting for their destination; they are eventually relocated by yard cranes and trailers in a more proper position. Thirdly, containers are loaded from blocks to trucks, and exit from the gate. Similarly, in a railway import cycle, yard cranes pick-up containers from blocks and load them on trailers moving from yard to the railway connection, where special cranes pick-up containers to put them on departing trains.

In a road/railway export cycle, the sequence goes in the opposite sense: from terminal gate or railway connection to blocks and then to vessel ships. In a transshipment cycle, when a vessel ship arrives, containers are downloaded to trailers by quay cranes, transferred to blocks by trailers, picked-up and stacked by yard cranes. After a consolidation and/or a delay in their position, containers are picked-up and transferred to the quay area, where they are loaded on a feeder ship. The opposite occurs if a feeder arrives and a vessel departs.

Here, the focus is on the first step of the import and transshipment cycles. Similar models can be defined for the following steps and the export cycle.

2.2 Agents in a Container Terminal

The main agents controlling a terminal are:

- the Container Agent (CA): each CA is an autonomous entity controlling the flow of a single or a group of containers (stowed in a ship bay or stacked in yard blocks);
- the *Quay crane Agent* (QA): it is an autonomous controller for a (set of) quay crane(s) with the same performances or the same reachable ship bays;
- the Trailer Agent (TA): it is associated to a (set of) trailer(s) with the same performances, or with the same reachable quay or yard spaces;
- the Yard crane Agent (YA): it manages a (set of) yard crane(s) with the same performances, or associated to the same yard blocks;
- the Railway crane Agent (RA): it controls a (set of) railway crane(s), to receive containers from trailers and load them on trains at the railway connection, or to deliver containers from trains to trailers;
- the *Truck Agent* (KA): it follows the operations executed by a (set of) truck(s), entering or leaving the terminal by its gate.

In import processes, the CA identifies the most suitable quay crane to download containers from ship, then it selects the trailers to transport containers to their assigned yard blocks, and finally the yard cranes to pick-up and stack containers in their assigned block positions. In export processes, the CA selects the yard cranes to pick-up containers from blocks, trailers to transport them to the quay area, and quay cranes to load them into their assigned bay-row-tier location in the ship.

The decisions taken by a CA are based upon real-time updated information received from agents of the alternative available cranes and trailers.

The global control of the activities in the terminal emerges from the behaviour of concurrently operating agents. The dynamical interaction between agents has to be analysed to specify the desired

system behaviour. The precedent observations lead us to examine interactions between a CA and several QAs, TAs, and YAs for downloading, transferring, and stacking containers, or for picking, transferring, and loading containers. Interactions exist also between a CA and YAs, TAs, and RAs/KAs when railway/road cycles are considered.

The agents' decisions are only limited by constraints of terminal spaces and resources (e.g., the limited number of quay cranes that can serve a fixed ship bay; the limited number of yard cranes), and of schedules for downloading/loading processes (establishing the number of containers moved for each ship bay, the location in the ship hold or cover, sequence of handling moves, the preferred crane).

Agents' decisions are fulfilled by human operators devoted to the associated resources (e.g. crane operators, trailer/truck drivers). The network of interacting agents may appear and behave as a unique distributed supervisor for the terminal.

3 DEVS MODELLING OF AGENTS' DYNAMICS

Each agent in the previously identified classes is described as an atomic DEVS. Namely, all agents interact by transmitting outputs and receiving inputs, which are all considered as event messages. Events are instantaneous, then timed activities are defined by a start-event and a stop-event.

For each agent, internal events are triggered by internal mechanisms, external input events (i.e. inputs) are determined by exogenous entities (other agents), and external output events (i.e. outputs) are generated and directed to other entities.

External or internal events change the agent state. An agent stays in a state until either it receives an input or the time specified before an internal event elapses. In the first case, an external transition function determines the state next to the received input; in the second case, an internal transition function gives the state next to the internal event. An output function generates the reactions of the agent.

The sequential state, \mathbf{s} , refers to the transition mechanism due to internal events, and is based on the current value of *status* (condition between consecutive events) and other information i peculiar to the considered agent: $\mathbf{s} = (status, i)$. The total state $\mathbf{q} = (\mathbf{s}, e, \mathbf{DL})$, where e is the time elapsed since the last transition, and \mathbf{DL} is the decision logic currently used by the agent to rank and choose the offers received by other agents.

For a CA, **s** may include information on: the current container position; the quay cranes available for negotiating handling operations; the trailers available for negotiating transport between quay and yard; the yard cranes available for negotiating handling operations from trailer to a block position or backwards; time scheduled in current state before the next internal event, if no input occurs.

For QAs, YAs, TAs, RAs, KAs, the sequential state may include the queued requests coming from CAs (for availability, for data about offered service, for confirmation of assigned service, etc.), and the time prospected before the next internal event.

To summarize, each agent can be represented as an atomic DEVS in the following way:

$$A = \langle \mathbf{X}, \mathbf{Y}, \mathbf{S}, \delta_{int}, \delta_{ext}, \lambda, ta \rangle$$
 (1)

where **X** is the set of inputs, **Y** is the set of outputs, **S** is the set of sequential states, $\delta_{int}: \mathbf{S} \rightarrow \mathbf{S}$ is the internal transition function, $\delta_{ext}: \mathbf{Q} \times \mathbf{X} \rightarrow \mathbf{S}$ is the external transition function, $\mathbf{Q} = \{\mathbf{q} = (\mathbf{s}, e, \mathbf{DL}) | \mathbf{s} \in \mathbf{S}, 0 \le e \le ta(\mathbf{s})\}$, $\lambda: \mathbf{S} \rightarrow \mathbf{Y}$ is the output function, $ta: \mathbf{S} \rightarrow \mathbf{R}_0^+$ is the time advance function, with \mathbf{R}_0^+ set of positive real numbers with 0 included.

The negotiation between agents is typically organized in the following steps. Announcement: an agent starts a bid and requires availability to other agents for a service. Offer: the agent requests data only to the available agents. Data regard the offered service the queried agents can guarantee. Reward: the agent selects the best offer between the collected replies and sends a rewarding message. Confirmation: the agent waits for a message from the rewarded agent, then it acquires the service. If confirmation does not arrive, then the agent selects another offer in the rank or restarts the bid.

In this paper, the focus is on the interactions between a CA with QAs, TAs and YAs for downloading containers from ship to a yard block. The study can be easily extended to other negotiations for loading or other processes. Besides, the *status* transitions are examined since the *status* is the main component of the sequential state.

3.1 Dynamics of Interactions between Agents in a Downloading Process

To download containers from a ship bay, transport and stack them into a yard block, each CA interacts with QAs to choose the quay crane for downloading, with TAs to select the trailers for moving the containers from the quay to the yard area, and finally with YAs to determine the yard cranes for stacking.

We assume that the CA firstly communicates exclusively with QAs, then with TAs only, finally with YAs. The negotiation is based on data about the offered services: a QA gives the estimated time to wait before the related crane can start downloading containers, and the estimated time to execute the task; a TA the estimated times to wait a trailer and for the transport task; a YA the estimated time for the yard crane to be ready close to the block, and the estimated time for the stacking task.

3.1.1 Interactions between a CA and QAs

For t<t_{C0} let the CA, say C, associated with a generic container be in a quiescent *status* (QUIESC) and let it begin its activity at t_{C0} (input X_{C0}). Then C spends the interval [t_{C0},t_{C1}] to send outputs Y_{C01}, Y_{C02},..., Y_{C0q} at instants t₀₁>t_{C0}, t₀₂,..., t_{0q}=t_{C1}. These messages request the availability to all the q alternative QAs of cranes that can serve the container. The sequence of requests (REQQAV) is not interrupted by any internal or external event. C makes transition at t_{C1} (internal event I_{C1}).

In $[t_{C1},t_{C2}]$ C waits for answers (WAIQAV) from QAs. Namely, the request C transmits to each QA may queue up with similar ones sent by other CAs. Next transition occurs at t_{C2} when either C receives all the answers from the queried QAs (X_{C1}) , or a specified time-out of WAIQAV expires before C receives all the answers. In case it receives no reply within the time-out (I_{C2}), C returns to REQQAV and repeats the request procedure. In case of time-out expiration and some replies received (I_{C3}), C considers only the received answers to proceed. The repeated lack of valid replies may occur for system congestion, crane failures, communication faults, or other unpredictable circumstances. To avoid indefinite circular waits and improve system faulttolerance, we use time-outs and let C repeat the cycle REQQAV-WAIQAV only a finite number of times, after which C is replaced by another agent.

If all or some replies arrive before the time-out expiration, C starts requesting service to the $g \le q$ available QAs at t_{C2} . In $[t_{C2},t_{C3}]$ C requests information with Y_{C11} , Y_{C12} , ..., Y_{C1g} at instants $t_{11} > t_{C2}$, t_{12} ,..., $t_{1g} = t_{C3}$. The uninterrupted sequence of requests in *status* REQQSE terminates at t_{C3} and C makes transition (I_{C4}).

Then, C spends $[t_{C3},t_{C4}]$ waiting for offers from the available QAs (WAIQOF), as the request C transmits to each QA may queue up with those sent by other CAs. Next transition occurs at t_{C4} when either all the answers from the QAs arrive (X_{C2}) or a time-out of WAIQOF expires. In case no reply comes within the time-out (I_{C5}) , C returns to REQQSE and restarts. If time-out expires and some replies are received (I_{C6}) , C considers only the

received offers. Again, cycling between REQQSE and WAIQOF is only for a finite number of times.

Once received the offers from QAs, C uses $[t_{C4},t_{C5}]$ to take a decision for selecting the quay crane (TAKQDE). At t_{C5} the decision algorithm ends (I_{C7}) : all the offers are ranked and a QA is selected.

Subsequently, C reserves the chosen crane with message Y_{C2} to the corresponding QA. C takes $[t_{C5},t_{C6}]$ for communicating the choice to the winner QA (COMCHQ). At t_{C6} the communication ends (I_{C8}) . Now, the QA sends X_{C5} : a rejection, if there is a conflict with another CA, or a confirmation. Hence, C uses $[t_{C6},t_{C7}]$ to wait for a confirmation from the selected QA (WAIQCO). The confirmation is necessary because CAs different from C act during the decision interval, and the selected crane can be no longer available. If C receives a rejection (X_{C3}) , or no reply within a time-out (I_{C9}) , it returns to COMCHQ, sends a new request of confirmation to the second QA in its rank. If C has no alternative destinations and the rejection (X_{C4}) or the time-out (I_{C10}) occurs, it returns to REQQAV and restarts.

At t_{C7} , after receiving a confirmation (X_{C5}) from the selected QA, C makes a transition to DWNLDG: interval [t_{C7} , t_{C8}] is for issuing the command Y_{C3} to the quay crane downloading the container.

Figure 1 depicts the complex dynamics previously described. Circles represent *status*-values and encapsulate outputs, arrows represent internal or input events.

3.1.2 Other Interactions

When, at time t_{C8} , the command is complete (I_{C11}), C starts the negotiation with TAs for a trailer.

C follows the same procedure as with QAs (requests and waits for availability and then for offers, decision for the best offering TA, waits for a confirmation/rejection). Then, the graph in figure 1 continues with a second part (not shown for lack of space), with the same structure as the first part. This guarantees modularity, which is important for simulation and control. After a confirmation, C issues a transport command, and the container is loaded on the vehicle associated to the selected TA.

Then, C starts the negotiation with YAs to stack the container in an assigned block position. Again, due to the modularity of the approach, the sequence of status-values follows the same protocol, and, finally, if a confirmation is received, C issues a stacking command and gets back to QUIESC.

C stops interacting and remains quiescent until the beginning of the next negotiation (if any) for downloading, consolidating or loading another container. The associated container is downloaded, transported and stacked in a block where it waits for the next destination (a block or a ship). If faults

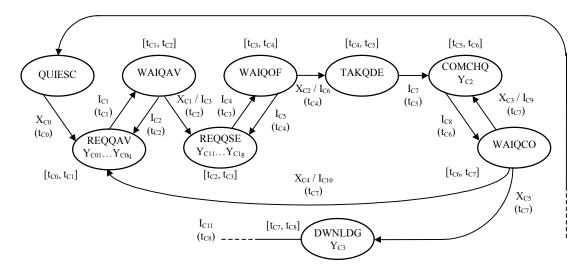


Figure 1: Dynamics of the interaction between a CA when negotiating with QAs.

occur to the selected cranes or trailer, *C* remains in QUIESC. Terminal operators restore the normal operating conditions and the container can be handled by the selected resources.

3.2 Specification of the DEVS Model of a Container Agent

We may identify the components of the DEVS model of a CA, on the basis of the negotiation mechanism previously described. Table 1 reports the admissible *status*-values. Detailed description of inputs, outputs and internal events is not reported for lack of space, but it can be easily derived from the interactions between a CA and QAs, TAs, and YAs.

The sequential state **s** collects the *status*-value and other information i about the CA, which is:

$$i = (p, AQ, AT, AY, ta(\mathbf{s}))$$
 (2)

including the current position p of the container (on ship, picked by quay crane, on trailer, picked by yard crane, in the yard block); the set AQ of quay cranes available for the currently negotiated downloading operations, or the set AT of trailers available for the currently negotiated transport, or the set AY of yard cranes available for currently negotiated stacking tasks; the time $ta(\mathbf{s})$ scheduled in current state before the next internal event.

The time advance function gives the residual time $ta(\mathbf{s})$ in state \mathbf{s} before the next scheduled internal invent. E.g., at the time t^* of entering a waiting status, the time-out fixes T_w , such that $ta(\mathbf{s}) = t^* + T_w - t$, where t is the current time.

Table 1: Status-values for a Container Agent.

Ct. t	1 1 1 1 D 1 1
Status	Activity Description
QUIESC	Agent quiescent
REQQAV	Request availability to all QAs
WAIQAV	Wait for availability from QAs
REQQSE	Request service to available QAs
WAIQOF	Wait for offers from available QAs
TAKQDE	Take decision for the best QA
COMCHQ	Communicate choice to selected QA
WAIQCO	Wait conf./reject. from selected QA
DWNLDG	Command selected QA to download
REQTAV	Request availability to all TAs
WAITAV	Wait for availability from TAs
REQTSE	Request service to available TAs
WAITOF	Wait for offers from available TAs
TAKTDE	Take decision for the best TA
COMCHT	Communicate choice to selected TA
WAITCO	Wait conf./reject. from selected TA
TRANSP	Command selected TA to transport
REQYAV	Request availability to all YAs
WAIYAV	Wait for availability from YAs
REQYSE	Request service to available YAs
WAIYOF	Wait for offers from available YAs
TAKYDE	Take decision for the best YA
COMCHY	Communicate choice to selected YA
WAIYCO	Wait conf./reject. from selected YA
STCKNG	Command selected YA to stack

DEVS models can be also specified for the other agents, and *status*-transition diagrams can be defined for the interactions occurring when agents negotiate tasks in loading or consolidation processes.

4 IDEAS FOR SIMULATING MAS

The DEVS atomic models can be integrated in a network, which can be used as a platform for simulating the MAS controlling a terminal, e.g. the Taranto Container Terminal (TCT).

In this context, it is possible to simulate not only the dynamics of terminal activities, the flow of containers, and the utilization of terminal resources (cranes, trailers, human operators, etc.), but also the efficiency of the MAS and its agents (flow of event messages, *status* transitions, waiting loops, etc.).

Then, two types of performance indices can be defined. Namely, it is possible to measure conventional indices: the total number of (imported, exported, transshipped) containers; the average throughput, during downloading (from ship to yard) or loading (from yard to ship) processes; the average lateness of containers in the terminal. Moreover, it is possible to measure the behaviour of the MAS and the efficiency of the agents' decision policies by means of: the average number of requests for each negotiation; the number of loops of *status*-values before a final decision is taken by a CA, expressed in percentile terms with respect to the total number of operations executed by every CA.

The performance measures can be evaluated both in steady-state operating conditions and in perturbed conditions. Perturbations may arise from: hardware faults or malfunctions; abrupt increase/decrease of maritime traffic volumes; sudden increase/reduction of yard space; traffic congestion of trailers; congestion, delays, message losses, and faults in the communication between agents.

Then, it is important to measure robustness of agents' decision laws, to see how they dynamically react to disturbances and parameter variations, and eventually to adapt them. The adaptation aims to make the autonomous agents learn the most appropriate decision laws in all terminal conditions.

To conclude, the simulation platform will allow to compare control architectures defined by:

- a static MAS in which CAs use heuristic decision parameters (estimated time of the requested task, distance of cranes or trailers);
- a dynamic MAS in which CAs take decisions by fuzzy weighted combinations of heuristic decision criteria; the weights can be adapted by an evolutionary genetic algorithm.

5 CONCLUSIONS

This paper proposes a MAS architecture for controlling operations in intermodal container terminal systems. The autonomous agents are represented as atomic DEVS components. The interactions between agents are modelled according to the DEVS formalism to represent negotiations for tasks when downloading containers from ship to yard stacking area. The developed model can be easily extended to describe other processes (loading containers from yard area to ships, redistributing containers in the yard area).

The DEVS model of the MAS can be used in a detailed simulation environment of the TCT, which allows to measure standard terminal performance indices and the efficiency of the MAS. Moreover, open issues are testing and comparing static MAS and dynamically adapted MAS, if evolutionary adaptation mechanisms are used.

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