IDEF method-based simulation model design and development

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Abstract: The purpose of this study is to provide an IDEF method-based integrated framework for a business process simulation model to reduce the model development time by increasing the communication and knowledge reusability during a simulation project. In this framework, simulation requirements are collected by a function modeling method (IDEF0) and a process modeling method (IDEF3). Based on these requirements, a common data model is constructed using the IDEF1X method. From this reusable data model, multiple simulation models are automatically generated using a database-driven simulation model development approach. The framework is claimed to help both requirement collection and experimentation phases during a simulation project by improving system knowledge, model reusability, and maintainability through the systematic use of three descriptive IDEF methods and the features of the relational database technologies. A complex semiconductor fabrication case study was used as a testbed to evaluate and illustrate the concepts and the framework. Two different simulation software products were used to develop and control the semiconductor model from the same knowledge base. The case study empirically showed that this framework could help improve the simulation project processes by using IDEF-based descriptive models and the relational database technology. Authors also concluded that this framework could be easily applied to other analytical model generation by separating the logic from the data.
Keywords: IDEF0, IDEF3, IDEF1X, discrete event simulation, business process

1 Introduction

Simulation is one of the most widely used decision aid tools due to its power, flexibility, and robustness. Particularly the discrete event simulation (DES) can model and analyze the behavior of many real life processes such as business processes, supply chain, and manufacturing processes. However, as Ryan et al. pointed out (2006), the simulation modeling often becomes a heavy programming task with the essence of the system being modeled lost in the detailed programming codes. In this way, the essence of the system is visible only to the code developers. This could create several potential problems for those who are involved in a simulation project. For example, it may create a serious information reusability problem. A simulation model is an abstracted representation of a real system to solve specific problems. Hence the information collected and extracted from the real system should be systematically represented and stored for future reuse in the form of systematic descriptions and formats. It may also cause a communication problem between developers and users. Typically users are domain experts who want to experiment with the simulation model to solve domain specific problems. This task requires frequent parameter changes and modification of the model. However, the heavy codes add difficulty to the proper management of this task. If we consider a simulation model development as a project, and if we have a structured systematic tool to support the simulation project, we believe that these problems could be managed. Sheppard (1983) proposed a widely cited “40-40-20” simulation model development time rule which states that analyst’s time should be distributed as follows for a successful simulation project: (1) 40% to requirement collection phase such as problem formulation, project planning, conceptual model development, and data collection; (2) 20% to model translation phase; (3) 40% to experimentation phase such as model verification, validation, implementation, and interpretation. Hence, for successful implementation of any simulation project, it is particularly important to have a right approach to the requirement collection and the experimentation phases. Hence, this paper intends to provide an integrated framework for those two phases in a simulation project.
The process description methods could play an important role in the simulation requirement collection phase. Although many process design, analysis and modeling (DAM) methods have been developed, using these methods in isolation – non-methodological approach – often fails to capture critical system behaviors due to the complexities and component interactions within the system. A methodological approach – systematic usage of a suite of methods – has a greater chance of success at representing critical system behaviors since it can account for diverse aspects of DAM activities such as information, function, and process interactions by a systematic and integrated usage of methods. IDEF (Integrated DEFinition) is a suite of descriptive modeling methods within which several different modeling languages are defined to describe systems from different perspectives. First, since IDEF is a well defined suite, it is considered to be easier to implement a methodological approach with the IDEF suite rather than with a completely different set of methods. Second since it is a descriptive modeling method, it could easily abstract and capture the essence of the system. In a typical simulation project, a project team consists of many team members such as system analysts, developers and domain experts. The system analysts collect and refine requirements with assistance from domain experts. This is an iterative communication process among all members. The ‘descriptiveness’ of IDEF methods could make this communication process easier and smoother than any other non-descriptive methods. For these reasons, IDEF methods have been a continued research subject.

The first category of the IDEF method related research attempted to build a generic and conceptual descriptive model using IDEF suites in a specific domain (Ang et al., 1994; Zhang et al., 1996). Another category proposed a way to generate an analytical model from a specific IDEF model. For example, an IDEF3 method has been used to generate simulation models using Witness simulation software (KBSI, 1995) and using Arena software (Resenburg et al., 1995). Jeong et al. (2008) developed a scheme to integrate the IDEF3 with a general open queuing network where IDEF3 works as a knowledge repository. The third category employed multiple IDEF methods and attempts to reuse common system knowledge among the different IDEF methods. For example, Lingzhi et al. (1996) proposed a scheme to integrate IDEF1 with IDEF0 for a computer integrated manufacturing information system design. Chen et al. (2004) also proposed a scheme to develop the enhanced IDEF1 information model based on the IDEF0-based process information,
which could serve as a base representation for an information model. This paper covers both the second and the third category together. It is an extension of Cho et al. (1999), KBSI (1995), and Chen et al. (2004) in that it attempts to provide an integrated framework of IDEF method-based simulation model design and development to help a successful simulation project.

![Conceptual framework of IDEF-based integrated approach](image)

Figure 1. "Conceptual framework of IDEF-based integrated approach"

The experimentation phase is an iterative process to perform model verification, validation, actual implementation, and interpretation that requires another 40% of analyst’s time. The verification, validation, and implementation processes frequently require significant code changes and model modification. It may also require multiple models with different parameter values and diverse scenarios. Hence, it would be helpful to automatically develop diverse versions of a simulation model from a common knowledge base developed through DAM activities. Although it is very difficult to develop a generic database-driven simulation model working for any domain, the domain specific database-driven simulation model development is feasible and useful. In fact, Pidd (1992) pointed out this fact by saying “generic simulators will not wholly replace simulation program for specific application.” In this paper, authors suggest a framework where a simulation model is automatically generated from the database obtained as a result of IDEF-based model design and development. Specifically, the functional system knowledge captured by IDEF0 and the process system knowledge captured by IDEF3 is used to develop and refine the data model knowledge captured by IDEF1X. Based on
IDEF1X, a simulation model is generated with a database-driven simulation model generation approach (DBSMGA). In this framework, the IDEF1X model works as a software independent knowledge base where different simulation software could access to generate models using their libraries. The framework is explained in Figure 1.

This framework is claimed to have the following advantages over non-methodological approaches:

- It facilitates knowledge reusability among different modeling methods during requirement collection phase.
- It can capture diverse aspects of real systems from different perspectives, which improve the accuracy of representation.
- It improves simulation model maintainability since the simulation logic can be changed inside the database as opposed to inside the simulation model.

However, it should be noted that authors do not find it necessary to use this framework for all situations. Instead, this paper claims that this framework has a better chance to lead a successful simulation project by improving communication and knowledge and model reusability. For example, simulation models may be directly built using an icon-based drag-and-drop approach if analysts have sufficient simulation software knowledge. However, based on authors’ experience, this direct model building practice without any methodological approach often generates incorrect models due to the lack of communication, improper model abstraction, and inappropriate model management techniques. Particularly, in a large-scale simulation project, it tends to add more difficulty to the proper simulation project management.

2 IDEF Methods for Business Process Knowledge Capture

According to Lin et al. (2002), a business process has the following elements:

- A business process has its customers.
- A business process is composed of activities whose objectives are to create values for customers.
• Activities are performed by actors who may be humans or machines

• A business process often involves organizational units which are responsible for the whole process.

We believe that IDEF methods well support those elements. For example, IDEF0 was designed to capture the ‘decisions’ and ‘activities’ of a system (Mayer et al., 1995). Those decisions and activities include information on what functions the system performs, what constraints the functions have, what is needed for functions, and what input and output are meaningful in performing those functions. An IDEF0 model is represented with rectangles with four different types of arrows surrounding the rectangles. A rectangle represents a function or activity described in a *verbal* phrase, and arrows represent (1) “Input” (on the left); (2) “Output” (on the right); (3) “Control” (on the top); and (4) “Mechanism” (on the bottom) called (ICOM) described in a *noun* phrase to explain the behavior of the function – see Figure 2 below. It also supports the hierarchical decomposition of activities for an appropriate abstraction of a system. We notice that the first three business elements could be supported by IDEF0. For example, IDEF0 model could be developed from a specific customer’s perspective and context – first element. The business activities are part of system activities – second element. The mechanism in ICOM includes actors – third element.

![Figure 2. "IDEF0 function modeling notation".](image)

IDEF3 is a process capture and description method within the context of a specific scenario (Mayer et al., 1995). The process schematic of IDEF3 has been widely accepted as a medium for process description in industry (Mo et al., 1998). As seen in Figure 3, the process schematic consists of the three main components: (1) “Unit Of Behavior” (UOB); (2) “Junction”; (3) “Link”.

![Figure 3. Process schematic of IDEF3.](image)
A UOB captures information on what is going on in a system to represent a process or an activity. It is depicted by a rectangle with a unique label. Junctions provide a mechanism specifying a logical branching of UOBs and introduce the timing – temporal – and sequencing of multiple processes. Junction types include an exclusive OR denoted by “X”, a conjunctive AND junction denoted by “&”, and an inclusive OR denoted by “O”. Since IDEF3 uses a scenario as a basic organization structure to describe how things work, it could easily fit with the first three elements of a business process. It also supports the top-down and bottom-up modeling sequences and hierarchical decomposition for multiple levels of abstraction. IDEF1X produces a data model that represents the structure and semantics of information within an enterprise or a system, known as business rules. An IDEF1X diagram is refined into three levels of detail: (1) an entity-relationship level; (2) a key-based level; (3) an attribute level. Diverse business rules are specified according to different levels of detail, and the model development is defined by these three level modeling procedures. Figure 4 shows summary of these three levels with symbols. Interested users are encouraged to read (KBSI, 1994) for detailed grammar and graphical symbols for the IDEF1X method.

Lin et al. (2002) also identified ten essential concepts useful in defining a business process, and we creatively used these concepts to investigate the fitness of the proposed IDEF method-based integrated framework for business process simulation model design and development. The results are summarized in Table 1. This table shows all but one concept are handled and represented by this framework.
At the same time, it also suggests the intervention from model developers and users are still necessary for verification and validation. In fact, authors do not claim that an automatic model development process could completely replace the insights and knowledge from humans - we even believe that it is not desirable. Instead, this integrated framework should be considered as an aid to support humans and their judgment.
3 IDEF1X Data Model Knowledge Base for Simulation Model

An IDEF1X data model should be robust enough to make the structural and semantic aspects of the simulation included in the model. One way to consider the semantic aspects of the simulation model is to study the simulation ontology, and reflect it within the data model design, since ontology provides the definition of the terminologies and relationship between them. Although several DES software packages are available in the market, they have some common basic structures or objects, while the unique structures are variations of these common structures. If a data model incorporates these structures into its design, it could be shared by different DES software packages.

Table 2 lists some of these basic common structures with their definitions within the context of business process simulation. Note that the neutral terminologies, Generator, Entity, Location, Resource, Queue, and Destroyer are used to avoid favoring a particular simulation language. For example, ED (2001) uses Source, Product, Server or Multiple-Service, Operator, Queue, and Sink while Flexsim (2007) uses Source, FlowItem, Processor or Multiprocessor, Operator, and Sink instead of these neutral terminologies.

<table>
<thead>
<tr>
<th>Name</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>Generator</td>
<td>A structure that creates entities to populate a model</td>
</tr>
<tr>
<td>Entity</td>
<td>A structure that flows through the model to represent customers, orders and any moving items in the model.</td>
</tr>
<tr>
<td>Location</td>
<td>A structure that interacts with an entity. This interaction is called a service, and it usually delays the progress of an entity through the model.</td>
</tr>
<tr>
<td>Resource</td>
<td>A structure that may be required by an entity or a location to provide a service. The difference between a location and a resource is that a location does not move, and a resource is moving toward a location when it is requested.</td>
</tr>
<tr>
<td>Queue</td>
<td>A structure that stores entities. The queue is awaiting service, not receiving service.</td>
</tr>
<tr>
<td>Destroyer</td>
<td>A structure that destroys entities</td>
</tr>
</tbody>
</table>

Table 2. “Basic simulation objects and definition”.

Figure 5 shows one possible mapping between an IDEF1X data model and corresponding neutral simulation structures (objects). The data model has multiple – one or more, denoted by P – Order/Product, Office/Shop, Employee/Equipment or Operator, and Storage/Queue objects. Each of which is mapped to Entity, Location, Resource and Queue object, respectively, in the simulation model. Note that since Generator and Destroyer are purely functional objects for an entity creation and destroy respectively, they do not need to be included in a data model.
Once a data model is built with consideration of simulation structures, the database can be easily created from which multiple simulation models could be automatically developed.

From a simulation perspective, the data model characterizes the required functionalities of simulation libraries. Hence if those functionalities are not supported by specific target simulation software, those functionalities should be developed to make the database-driven simulation model development easier.

4 Database-Driven Simulation Model Development

Most simulation software provides its own script language with structured query language (SQL) and Open Database Connectivity (ODBC) capability. Hence, the interface between the simulation and database management system (DBMS) becomes easier but it still requires some coding efforts. In this study, both Flexsim and ED libraries were adopted to implement DBSMGA since both provide the Object-Oriented customized library development capability in addition to the rich standard libraries. The customized library development capability is supported using Flexscript (Flexsim’s script language) or C/C++ in case of Flexsim and 4D Scripts in case of ED. Each object (library) consists of a set of attributes (characteristics) and methods (functions) that can be implemented when an associated event handler is activated in the library. In case of Flexsim, some typical examples of an event handler are OnReset – triggered when users click the
reset button, OnMessage – triggered when an object receives a message from other objects, OnEntry – triggered when an entity enters current object, and OnExit – triggered when an entity leaves from a current object. All properties of a class are inherited to the instances of the class when they are created in a model. By combining the standard library and the customized library, users can develop their own user-specific and/or domain-specific simulation software, which makes mapping from the data model to the simulation easier.

In the case study to be discussed later, we first developed a Flexsim model. A customized library named “simulation-generator” was developed to generate simulation models from a database. Note that this corresponds to the Simulation Model Generator in Figure 1. All codes were written in the User event handler and the total lines of the codes are less than 250 including all user-interface and screen embellishment.

<table>
<thead>
<tr>
<th>NO</th>
<th>Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>dbopen(&quot;MyDSN&quot;,&quot;select * from Equipment&quot;,0);</td>
<td>Connected to the database table Equipment through the DSN defined at MyDSN and perform SQL statement.</td>
</tr>
<tr>
<td>2</td>
<td>settablenum(&quot;InfoTb&quot;,1,1,dbgetnumrows());</td>
<td>Store total number of records in database at cell(1,1) in InfoTb. &quot;dbgetnumrows()&quot; is a key word to count the number of records in the current database.</td>
</tr>
<tr>
<td>3</td>
<td>for(int i = 1; i &lt;= dbgetnumrows(); i++) {</td>
<td>Repeat the function defined at { } dbgetnumrows() times.</td>
</tr>
<tr>
<td>4</td>
<td>createinstance(node(&quot;/SHOP-Equipment&quot;,library()),model());</td>
<td>Creates an instance of the SHOP-Equipment class (library) and places it in the model. Note that SHOP-Equipment is a customized library.</td>
</tr>
<tr>
<td>5</td>
<td>setname(last(model()),dbgettablecell(i,1));</td>
<td>Change the name of an instance using the name stored at the Equipment database. dbgettablecell(i,1) reads the string data stored at ith row and 1st column</td>
</tr>
<tr>
<td>6</td>
<td>createinstance(node(&quot;/SHOP-Buffer&quot;,library()),model());</td>
<td>Creates an instance of the SHOP-Buffer class (library) and places it in the model. This serves as a queue for equipment.</td>
</tr>
<tr>
<td>7</td>
<td>setname(last(model()),concat(dbgettablecell(i,1),&quot;Q&quot;));</td>
<td>Define the name of the queue in front of equipment. Concat connects multiple strings.</td>
</tr>
<tr>
<td>8</td>
<td>contextdragconnection(prev(last(model())),last(model()), &quot;A&quot;);</td>
<td>Connect the output port of queue to the input port of the equipment</td>
</tr>
<tr>
<td>9</td>
<td>Set_Equipment_Attributes;</td>
<td>set equipment attributes (user-defined function)</td>
</tr>
<tr>
<td>10</td>
<td>SetQueue_Attributes);</td>
<td>set queue attributes (user-defined function) and end of For statement</td>
</tr>
<tr>
<td>11</td>
<td>Dbclose();</td>
<td>Close equipment database</td>
</tr>
</tbody>
</table>

Table 3. “Partial pseudo code in simulation model generator”.

Table 3 shows the partial pseudo-codes to create instances of a library object from Equipment table to explain how to read the database through ODBC (lines 1 and 10) and create objects (instances of a library) and connect them using the ports.
One record in *Equipment* table represents one machine instance whose field consists of (equipment ID, equipment Name, Capacity, MTBF, MTTR etc). Lines 8 and 9 are calling user-defined functions to define the attributes of all instances. It is important to recognize that a model is a set of instances of all classes (or objects) located in the library. These classes may be provided by vendors or newly created by users. In this study, the “SHOP-Equipment” object and “SHOP-Buffer” object were created in the library by authors using Flexscript to facilitate the mapping from the data model to Flexsim.

5 Guidelines for Knowledge Reusability among IDEF methods

Developing a descriptive model using IDEF methods requires a feedback loop for obtaining consensus and confirmation from domain experts. Considering the fact that each IDEF uses a different modeling language to capture different perspectives of the real systems, the captured knowledge reuse among different IDEF methods has proven difficult to generalize. However, the importance of the knowledge reuse is critical – remember 40-40-20 rule, and how much time is required for requirement collection and experimentation? Based on our experience, the following guidelines seem to be useful in the knowledge reuse among IDEF0, IDEF3 and IDEF1X. As stated previously, IDEF0 represents the functional behavior of a system through four different types of data (ICOM) and a set of activities. The ICOM data could be information, objects or anything described in a *noun* phrase. Hence, some of the IDEF0 data may be represented as an object or an attribute in the IDEF1X model. Although IDEF0 is not designed to capture the temporal relationship among activities, some functional aspects of the system may include the temporal relationship among activities. Hence, if this happens, some activities in IDEF0 could be also represented in the IDEF3 diagram. Guidelines recommended for knowledge reusability among IDEF methods include:

- An IDEF0 modeling is recommended first since it provides overall system level knowledge.

- Then, an IDEF3 modeling is recommended if temporal information among activities is needed. Note that some descriptions in UOBs may provide a clue for attributes and objectives in an IDEF1X model.
• With help of IDEF0 and IDEF3, the detailed IDEF1X model could be developed.

• A “Mechanism” in IDEF0 may easily turn out to be an object in IDEF1X since “Mechanism” often includes resources – a typical object in an IDEF1X model.

• The “Constraint” in IDEF0 may provide a logical object in IDEF1X since this “Constraint” often represents the business rules of a system.

• Check if there is any noun in the verbal phrase (function) in an IDEF0 model that needs to be translated into an attribute or object in an IDEF1X model. Since a verbal phrase explains the behavior of the function, some nouns used in the phrase may convey meaningful information for a data model.

6 Case Study

This paper employs a case study of a semiconductor fabrication process to illustrate the concepts and framework stated in this paper. The IDEF1X data model created for this case study could also apply to many real-life business process problems. This case study originally came from Deuermeyer et al. (1993), and it was adopted here since it involves very complex real-life business processes. This case analyzes 172-step semiconductor wafer fabrication processes with six work-areas - CLEAN, STRIP, IMPLANT, DEPOSIT, LITHO and ETCH. These areas perform wafer cleaning, stripping, ion implantation, deposition, lithography and etch operation, respectively. Each work-area consists of machines and operators. An operator is required for wafer-loading and unloading operations at each machine. In addition, when different wafer-lots are loaded, the set up is required. The different operations may have different processing times even though they are performed in the same machine. Each product type can have its own routing as in a general job shop. However, most operation sequences are similar across product types in the wafer fabrication process. For example, all wafers start their operations from CLEAN area and finish at DEPOSIT area after several intermediate operations. One of the typical operation sequences is cleaning, litho, implantation, striping, deposition, etching, striping and deposition again. The important shop information such as the number of machines and operators, MTBF (mean time between failure), and MTTR (mean time to repair) is summarized in Table 4.
that both MTBF and MTTR are exponentially distributed, denoted by expo(). Two virtual areas were added to indicate the starting (START) and ending (END) of the process. Note that the second column consists of 8-tuples and each of which denotes the number of visits to each work-area. This 8-tuples are arranged in the orders of START, CLEAN, STRIP, IMPLANT, DEPOSIT, LITHO, ETCH and END. It is assumed that this shop is operating for 24 hours with 3 eight-hour shifts due to high capital equipment. The main performance measure for this shop is the system cycle time.

<table>
<thead>
<tr>
<th>Work Area</th>
<th># of Visits to Work-Area</th>
<th>No. of Machines</th>
<th>No. of Operators</th>
<th>MTBF (hrs)</th>
<th>MTTR (hrs)</th>
</tr>
</thead>
<tbody>
<tr>
<td>START</td>
<td>(0, 1, 0, 0, 0, 0, 0, 0)</td>
<td>4</td>
<td>1</td>
<td>expo(42.18)</td>
<td>expo(2.2)</td>
</tr>
<tr>
<td>CLEAN</td>
<td>(0, 3, 0, 0, 15, 1, 0, 0)</td>
<td>3</td>
<td>1</td>
<td>expo(55.18)</td>
<td>expo(12.86)</td>
</tr>
<tr>
<td>STRIP</td>
<td>(0, 2, 0, 0, 1, 1, 9, 0)</td>
<td>3</td>
<td>1</td>
<td>expo(75.93)</td>
<td>expo(3.88)</td>
</tr>
<tr>
<td>IMPLANT</td>
<td>(0, 1, 6, 1, 0, 0, 0, 0)</td>
<td>5</td>
<td>1</td>
<td>expo(100)</td>
<td>expo(2.78)</td>
</tr>
<tr>
<td>DEPOSIT</td>
<td>(0, 2, 0, 0, 8, 9, 8, 1)</td>
<td>20</td>
<td>3</td>
<td>expo(62.91)</td>
<td>expo(9.35)</td>
</tr>
<tr>
<td>LITHO</td>
<td>(0, 5, 12, 7, 2, 33, 6, 0)</td>
<td>33</td>
<td>4</td>
<td>expo(100)</td>
<td>expo(2.78)</td>
</tr>
<tr>
<td>END</td>
<td>(0, 0, 0, 0, 0, 0, 0, 0)</td>
<td>28</td>
<td>3</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table 4. "Facility data by work-area".

Based on the description above, we attempt to build an IDEF0 model according to the first guideline in the previous section. Each area could be modeled as an activity in IDEF0, and these activities are connected with each other through wafers. For example, Figure 6 shows part of the IDEF0 model in the wafer strip area with its decomposition to show the wafer strip process in detail. For any machine, when a batch of wafers (job) arrives, the operators are responsible for selecting the proper job according to the pre-determined dispatching rule that decides the job processing sequence at the machine. Once a job is selected, it is loaded onto a machine, and the set up occurs if the job’s lot number is different from that of the previous job. Once it finishes its operation, the wafers are unloaded and are ready to move to the next destination. The wafer changes its status over time as seen in the figure. For each area, these wafer processes are repeated.
Next, according to the second guideline, we tried to consider building an IDEF3 process model. All 172 steps are needed to be represented by IDEF3 model. In this case, the IDEF3 model is same as the process plan of a wafer fabrication, whose routing was already depicted in Deuermeyer et al. (1993). Hence, it was not repeated here. According to the fourth guideline, both an operator and a machine – mechanism – are considered as important objects, and they are incorporated into the IDEF1X data model as an object. According to the fifth guideline, the constraint information, Dispatching Rule, is also captured within the data model since it is considered as an important factor affecting the cycle time from the shop scheduling perspective. The same is true for the lot size (Transfer Batch Size) constraint information. It is also observed that the several functions are controlled by a
constraint called *Machine Operation Recipe*, which may contain the operation sequence information for each product.

With the help of IDEF0 model and the problem descriptions, the IDEF1X model is defined as in Figure 7. Based on above discussions associated with the IDEF0 model, the *Equipment, Product, Operator* and *DispatchRule* are first defined as an independent object (entity) with keys and attributes. For example, each equipment needs *capacity, MTBF, MTTR, setup time and run time, WorkingShift, Overtime, BufferSize* and *Dispatching_ID* information. The *BufferSize* defines the size of buffer in front of an equipment holding parts awaiting processing, and the *Dispatching_ID* is the set of rules defining the sequence of jobs in the queue. Typically, it follows FIFO – First-In-First Out. In this case study, the queue object represented in Table 2 is not handled as a separate object since all queues in front of each piece of equipment are considered as infinite.

![IDEF1X data model for case study](image)

The *Product* object can represent a bill-of-material (BOM) information through *Quantity in Parent* and *Parent ID* attributes. The *demand* and *LotSize* are attributes that affect the cycle time. The *Operation* object is defined as a dependent object since it can be uniquely identified only through *Equipment, Product* and *Operator* objects. Hence, the relationship between these three objects and an *Operation* is an identifying connection with one-to-many (one or more) cardinality. However, the relationship between *DispatchingRule* and *Equipment* is a non-identifying with one-to-many (at least one) cardinality since *DispatchingRule_ID* is used as a non-primary foreign key in the *Equipment*.  

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It is important to understand the difference between the Operation and the Routing object. The Operation contains all operation information that describes “who (operator) handles what products with what machines for what time”, and the Operation_Code can be used as an alternative primary key. The Routing object is created for simulation model generation to describe the sequence of operations for each product using the information in an Operation object. For each product type (non-primary FK), the source operation code (OperationCode_From) and the destination operation code (OperationCode_To) is described with its corresponding routing probability (Percentage) to support the probabilistic routing view. The Routing object provides the sequence of operations for each product type while these operations are characterized by the Operation object. The prototype data model in Figure 7 was translated into the corresponding MS-ACCESS™ database using the SmartER™ case tool developed by KBSI (1994). Figure 8 shows a snapshot of the Flexsim simulation model generated from the data model in Figure 7 using the codes in “simulation-generator” library whose partial pseudo-codes are represented in Table 3.

![Image](https://example.com/image.png)

Figure 8. “A snapshot of Flexsim model from database”.

In this figure, each of six work-areas is described in bold while the name of an object is represented in a regular letter. When multiple Operators are involved, all operators are directly connected to the Dispatcher object which is directly controlled by a Processor object. Figure 9 shows that of the ED simulation model from the same data model. The library object in ED is called an atom. The queue atom is connected to the operation atom, which is connected to the routing atom that connects the work-areas. The operator control atom (OP CONTROL) is connected to both the operator atom and the operation atom. The first atom
denoted by "1" represents a product atom corresponding to the Product object in Figure 7. The queue atom denoted by "2" has infinite capacity, which corresponds to a Buffer object. The operation atom denoted by "3" contains all equipment information such as MTBF and MTTR as its attributes, and it also includes all sub-operations such as loading, set up, cleaning operation and unloading operation.

When a sub-operation requires an operator, the atom performing the operation (i.e. cleaning atom) sends an operator-request-message (ORM) to a corresponding operator control atom (OP CONTROL). The operator control atom matches an ORM to an available operator, and it sends available operator(s) to the requesting atom. If there is no available operator for that ORM, it has to wait at the internal message queue inside an operator control atom. Once the (sub) operation finishes, the operator is released from the requesting atom and it becomes available again. All channels are connected using the information in the Routing object, and sub-operation information in an operation atom comes from the Operation object in Figure 7. The operator atom corresponds to the Operator object.

This model was executed with the data in Table 4 for 5 times to filter variation, considered as the first alternative (ALT1). Each run has 60,000 simulation hours after 10,000 hours warm up period. Since the sum of the three operators’ utilization in CLEAN, STRIP and IMPLANT was around 80 %, we created two other alternatives where these three areas have two shared operators (ALT2) and one shared operator (ALT3). The corresponding models were quickly generated again
from the database by changing the "Operator_ID" in the Operation table without modifying any logic in Flexsim environment for different scenarios. The average cycle time and throughput (number of units produced per day) were compared and displayed in Figure 10 where the bar shows the average cycle time and the line represents the throughput. As seen in the figure, the performance of the second alternative (ALT2) is almost identical to that of the first alternative (ALT1) even with less number of operators, and both outperform the third alternative (ALT3).

![Figure 10. "Alternative comparison".](image)

Through this case study, we showed that the IDEF method-based integrated framework could help improve the process of a simulation project by using IDEF-based descriptive models to capture requirements and to perform the experimentation. The IDEF1X-based data model supported by IDEF0 and IDEF3 could reduce the time and effort for simulation model development and maintenance. Before closing this section, it should be recognized again that the DBSMGA does not depend on any specific simulation software. Any simulation software supporting the ODBC and SQL capabilities could be used. If the software has the capability to customize the standard library, it could also reduce the effort required to map the data model into the simulation model.

7 Discussion and Conclusion

In this paper, the integrated framework of IDEF method-based simulation model design and development was provided for a business process. In this framework, the systematic use of IDEF0 and IDEF3 for business processes was proposed to help the requirement collection phase in a simulation project. From this systematic use of both descriptive models, the IDEF1X-based data model was created and became a knowledge base from which multiple simulation models could be developed, which could save time and effort in the experimentation phase in a
simulation project. A case study in a semiconductor manufacturer was conducted to show the feasibility of this framework where both Flexsim model and ED model were generated. This paper also discussed the guidelines to reuse the captured system knowledge among IDEF0, IDEF3 and IDEF1X.

The advantages of this integrated framework are to improve design knowledge reusability among IDEF0, IDEF3 and IDEF1X. It could also significantly reduce simulation model development and maintenance effort. By combining both IDEF methods and the database technologies together, this research significantly improved the previous IDEF based researches in that this framework provided a specific, systematic way to implement and execute the previous IDEF based modeling and design works. Many practitioners and simulation developers have been using the icon-based graphic user interface, and they should be familiar with all icons to develop and use the simulation models – This naturally leads to more focus on the model development phase without the ‘descriptiveness’ for better communication. However, the framework used here could change this game rule. The use of IDEF methods leads to more focus on the requirement collection phase of the simulation project. Also by using database as a knowledge base, this framework eliminated the dependence on the specific simulation software, and increased the efficiency in the experimentation phase of a simulation project. Authors believe that the results will significantly contribute to the successful use of simulation in the business process area where requirement collection is considered most difficult but important. Another direct advantage of this framework is that this could be applied to any analytical model as long as that model supports the database technology by separating the logic from the data.

The result of this study could provide many new ideas and suggestions to both practitioners and researchers. We summarized these into two categories: IDEF method-based modeling category and the database-based model generation category. Regarding the first category, it would be very useful to automate the knowledge conversion mechanism among IDEF methods to support human judgment and communication during the large scale simulation project. Also although we have provided many rules and insights for this conversion, more research is expected to enrich these lists in a specific domain and/or generic domain. In addition to the IDEF methods, the unified modeling language (UML) could also be considered for this DAM activities since it also supports the diverse modeling approaches from different perspectives. Designing more suitable data
model and/or database model could be another direct research area in this category since there may be an optimal data model in a specific domain. In the second category, the direct research area includes the development of an integrated simulation model generator through which the same knowledge in the database is transformed into software specific simulation models. In this way, the knowledge reusability will be maximized among simulation software products. Also, integrating all these – both descriptive modeling methods and the database model – within single platform could also be considered as another promising future research area.

References


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