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Toward an Aggregate Approach for Supporting Adaptive Sales And Operations Planning

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Abstract—The Demand-Driven Adaptive Enterprise (DDAE) model introduced by the Demand Driven Institute (DDI) few years ago becomes of prime interest for both scholars and practitioners. This research work is investigating the rarely studied strategic part of this DDAE model called Adaptive Sales and Operations Planning (AS&OP) process. One of the main issues regarding this strategic process is to determine how to model it through product family aggregates. Actually, literature analysis demonstrated that no solution exists to support such a process. This research work intends to solve this issue by designing a first AS&OP model allowing an aggregate reasoning. This proposal has been successfully tested on a illustrative but realistic example.

Keywords—component; S&OP; AS&OP; DDAE; DDMRP; aggregate production planning; decision support system

I. INTRODUCTION

Market globalization and industrial competition have led companies to maximize their profitability while better considering uncertainties and variabilities. Regarding Supply Chain Management (SCM) practices, experts and scholars have developed methods and techniques able to handle the stakes of profitability such as JIT, TOC, KANBAN, MRP2, ... [1]. The objective is systematically to respond to the customers' demand while reducing costs by optimizing resources' utilization. But as demonstrated by [2], most of these methods failed to manage properly the increasing uncertainties and variabilities that companies have to cope with nowadays. In a context where supply chain disruptions are becoming the norm, a recent innovation stands out: the Demand Driven Adaptive Enterprise (DDAE) [3].

Since 2011, DDAE attempts to make a synthesis of good practices from previous material management methods and adds some valuable innovations [3]. It has shown its ability to be agile and robust in disrupted context when compared to Kanban and MRP2 [2]. However, despite the increasing interest from both practitioners and scholars for DDAE, it is still in its infancy and needs additional contributions to consolidate both the method itself and the associated decision support systems [2]. The Demand Driven Institute (DDI) spent a lot of effort during the last decade on the operating model (Demand Driven Operating Model or Demand-Driven Material Requirement Planning) while staying more vague on the tactical and strategical levels of DDAE. In particular, the strategic layer of DDAE, called

Adaptive Sales and Operations Planning (AS&OP) has rarely been studied and, would benefit a lot from additional studies to support concrete implementations. Among all potential improvements, one key issue is about the granularity level that should be used to complete the AS&OP decision making. Ones indicate that it should be managed by aggregating elementary calculations made at the product (variant A) level while others indicate that the process must directly be modeled as aggregated (through families of products) (Variant B) to support the analysis and the associated decisions. For long time, it has been demonstrated that Variant B allows major benefits such as minimizing staffing fluctuations, reducing overheads or increasing production efficiency [4]-[6]. However, it has also been demonstrated that such a Variant is not trivial to be implemented in real context [7]-[9].

This research work focuses on this problem statement and attempts (i) to demonstrate the feasibility of using an aggregated model (Variant B) to support the AS&OP process and (ii) to discuss benefits and limits of following such an approach.

The remainder of the paper is composed of 4 main sections. First, a literature analysis is developed in order to position the research question and the expected contribution. Second, an aggregated AS&OP model is proposed for supporting decision making. Third, an illustrative case is depicted in order to validate the proposal and to discuss its benefits and limits. Fourth, a quick conclusion is given and avenues for future research are described.

II. BACKGROUND AND PROBLEM STATEMENT

A. DDAE Model

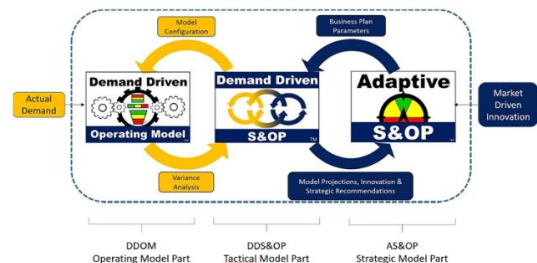


Figure 1. DDAE model [10].

Since 2011, the Demand Driven Institute (DDI) has designed, developed and popularized the DDAE model through trainings, conferences, implementations and publications [1],[3],[10],[11]. Basically, the DDAE model follows a classic planning hierarchical structure (figure 1.), as for MRP2 [12]. Typically, it includes three planning levels dedicated to respectively operational (Demand Driven Operating Model DDOM), tactical (Demand Driven Sales and Operations Planning DDS&OP) and strategic (Adaptive Sales and Operations Planning AS&OP) issues. The basic mechanics of the DDOM consists in identifying buffered items and replenish buffers (figure 2.) to a top of green zone anytime they (similar to inventory position) falls into a green zone. One of the specificities of DDOM is that Green and Yellow zones change over time according to a forecasted Average Daily Usage. Another one is to deal differently with usual and spike buffer consumption.

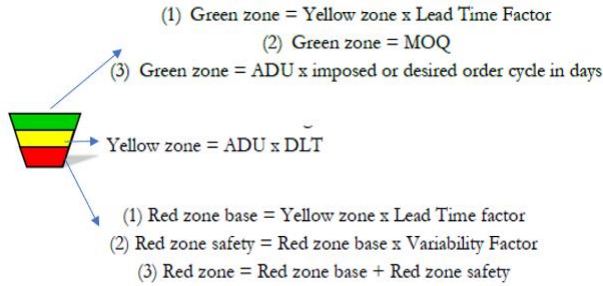


Figure 2. Principle of a stock buffer in DDOM [13].

For full detail on the DDAE mechanism, the reader can refer to [1],[3],[10],[11].

More recently, scholars have started to study the DDAE paradigm [2],[13]-[16] notably by focusing on the DDOM, previously called Demand Driven Materials Requirement Planning (DDMRP). They demonstrated ([2],[13]-[16]) the agility of DDMRP, especially in disrupted and uncertain environments, for maintaining performances (costs, lead times and inventories). DDMRP model behaviors have also been exhibited by studying its variables and by challenging the empirical elements of the model. Martin et al., 2018 [11], for instance, introduced a formal process modeling of DDS&OP and AS&OP.

Martin et al., 2018 [11] underlined that AS&OP features are very close to the usual S&OP ones but highlighted some specificities: buffer positioning at strategic level, capability to deal with strategic opportunities. Vidal et al., 2018 [17] made an empirical comparison of S&OP and AS&OP. They show, among others, that an ambiguity remains regarding the granularity level that has to be considered to support the AS&OP process. The DDI ([3], [10]) indicates that, as for usual S&OP best practices [18]-[20],[22]-[24], the process should consider aggregates as a reference to make calculations and analyses. Because of the simplicity of the DDAE order generation process, the accessibility to DDOM simulation and management software, practitioners work directly with elementary end items as the DDAE model does

not include a clear and formal method to do it differently. In this paper, aggregated products are defined as a Family: *A group of end items whose similarity of design and manufacture facilitates their being planned in aggregate, whose sales performance is monitored together, and (occasionally) whose cost is aggregated at this level* [21].

One of the strengths of the DDAE model is to focus on the use of relevant data at each level of decision-making. Typically, product sales forecasts at the operational level are used for designing buffers but not for order generation. For AS&OP, DDI trainings and publications[1] recommend to respect this practice of using relevant data, but no proposal for modeling operations at this level is given. In addition, the few examples on the DDS&OP subject aggregate data (Variant A) what seems to go against the best practices on the subject (Variant B)

B. Research Question

Based on the previous statements, the following research question can be formulated: *Is it possible to support an AS&OP process through a pure aggregated perspective? If yes, how this approach can be technically supported?*

The objective is finally to develop a solution that consider only families of products at the strategic level (Variant B) instead of a simple aggregation of elements resulting from the simulation of the operational DDOM to elementary end items (Variant A).

The following section will develop such a model.

III. PROPOSAL

A. Notation

Table I defines the notations used in this research work.

TABLE I. NOTATIONS

Notations	
General notation	Specific notation from DDAE
F_i : Product Family i	SB ij : Stock Buffer of Product j of Family i
FC m : Component Family m	ADU ij : Average Daily Usage of Product j of Family i
FPC p : Purchased Family p	DLT ij : Decoupled Lead Time of Product j of Family i
P ij : Product j of Family i	MOQ ij : Minimum Order Quantity of Product j of Family i
C mn : Component n of Family m	WOC r : Work Center r
PC pq : Purchased Component q of Purchased Family p	WOCrit : Critical Work Center
WOC r : Work Center r	capaWOCrit : Capability of WOCrit
WH : Warehouse	Y Zij : Yellow Zone of Product j of Family i
α_{ij} : Mixte Product, % P ij in family i	R Zij : Red Zone of Product j of Family i
β_{Fi} : Family mix, % F_i in total	LTF ij : Lead Time Factor of Product j of Family i
θ_{Fi} : Family mix i of	V Fij : Variability Factor of Product j of Family i
ADU cri : %contribution F_i in total	OC ij : Order Cycle of Product j of Family i
BOM jm : Bill of Material of P ij and component m	SB Fi , ADU i , DLT i , G Zi , Y Zi , R Zi , LTF i same notions to aggregate (family product level i).
Av li or k : Average Inventory of Family i OR component k	
PR j : Aggregate Price of Product of Family i	
PRC m : Aggregate Price of Component of Family m	We can substitute ij by C mn or PC pq .

Notations	
General notation	Specific notation from DDAE
PRPCp : Aggregate Price of Purchased Component of Family p CMi : Contribution Margin of one product of Family WK : Working Capital CM : Contribution Margin	i by Cm or PCp in order to obtain correct notation necessary about component and purchased component (exemple : ADUij give ADUCmn, ADUPCpq) DAFi: Demand Adjustment Factor of ADUi ADUCrit: Average Daily Usage Aggregate Family Product or Component on the flow of WCri

B. Assumptions

To develop an aggregate planning approach, the following assumptions have been made:

- (A1) Be able to provide an average daily usage ADUi to the product family reference over the required horizon;
- (A2) Be able to provide α_{ij} , β_{Fi} and calculate θ_{Fi} ;
- (A3) All considered buffers are managed within a DDOM approach;
- (A4) Product families are elaborated with a production perspective (same production process within a family);
- (A5) buffers' location are the same for all items of a given family.
- (A6) DLTij and LTFij are similar for any item j of a same family Fi..
- (A7) FCm families are built as Fi families, i.e. considering workflow similarities and FCm as a finished product. FPCp families are constituted by supplier.
- (A8) Critical resources are known.
- (A9) Spikes' management is not considered in this version of the model.

C. Aggregate DDOM model at product family level

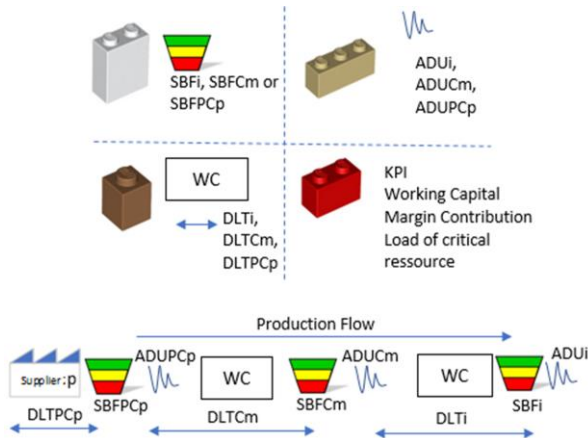


Figure 3. Four Elementary Brick of an Aggregate model at product family in DDOM environment.

Our main objective is to develop generic DDOM elementary bricks at the product family level in order to be able to build a specific “bricks assembly” for each industrial situation. The following three-step approach (figure 3.) has been set up to reach this goal:

- (1) Define a SBFi model of a buffer aggregated to the any family Fi;
- (2) Determine how ADUi is propagated, calculate ADUCm and ADUPCp and then derive from (1) and (2) the possibility of managing SBFCm and SBFCp like SBFi;
- (3) Calculate strategic Key Performance Indicators (KPI) recommended by the DDAE model: Working Capital, Margin Contribution and Load Ratio of Critical Resources.

In the following, each step of the proposed method is described in detail.

1) Step1: Family Buffer SBFi definition

At the DDOM level, each buffer SBij is linked to two groups of data: (i) Individual Part Properties and (ii) generic Buffer Profiles (Figure 4.).

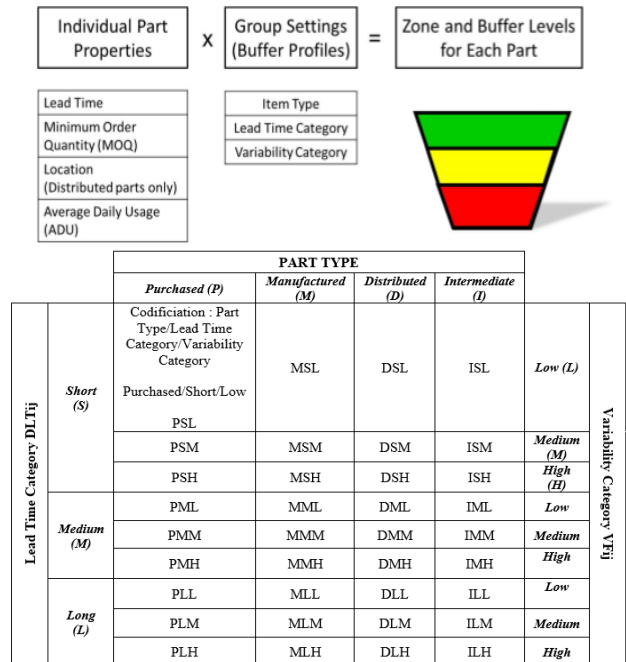


Figure 4. Buffer Profiles and Individual Part Properties of Stock Buffer[13].

Each buffer SBij is therefore specific to its product Pij. To have information usable at the AS&OP level, we could aggregate this specific information to obtain the data useful for quantifying the key performance indicators. But ADUij knowledge is not relevant over a long projection horizon. ADUi information being relevant, we propose to include it as basic data of our SBFi brick. In order to develop our AS&OP Model (Variant B), we propose to integrate two elements at the level of the basic calculation formulas (figure 2.) for the parts of an SBFij in DDOM Model:

- (1) The characteristics LTFij and DLTij are similar to the family i (Assumption (A6)) denoted LTFi and DLTi.
- (2) Each Pij has its own ADUij (example: a seasonal product within a family product) and it is desirable to keep the impact of this at the aggregate level. By integrating the mixed predictive product α_{ij} , we can

from ADU_i (Assumption (A1)(A6)) reach an image of the impact of each component of the family via its own buffer characteristics at the desired level. We use $ADU_{ij} = \alpha_{ij} \times ADU_i$.

Concerning the green zone GZ_i, a partial aggregation results of assumption (A1) on ADU_i but a transition to the product reference via α_{ij} takes into account specificities such as Minimal Order Quantity MOQ_{ij} or OC_{ij}.

Concerning the tallow zone YZ_i, an aggregation to the family of products with the use of DLT_i and ADU_i (assumption (A2)).

Concerning red zones RZ_i, a partial aggregation results of assumption (A1) and the use of ADU_i and LTF_i but a transition to the product reference via α_{ij} to take into account the impact of variability VF_{ij}.

Our proposal is summarized in figure 5. and constitutes a tradeoff allowing an aggregate model.

	DDOM Model	AS&OP Model
	Product Buffer P _{ij}	Family Product Buffer associated F _i
Green Zone	(1) MOQ _{ij} (2) YZ _{ij} x LTF _{ij} (3) ADU _{ij} x OC _{ij} GZ _{ij} = MAX(1;2;3)	ADU _{ij} = $\alpha_{ij} \times ADU_i$ $GZ_i = \sum_j GZ_{ij}$
Yellow Zone	YZ _{ij} = ADU _{ij} x DLT _{ij}	YZ _i = ADU _i x DLT _i
Red Zone	YZ _{ij} x LTF _{ij} x (1 + VF _{ij})	RZ _i = YZ _i x LTF _i x (1 + $\sum_j \alpha_{ij} \cdot VF_{ij}$)

Figure 5. Characterization of a Aggregate Stock Buffer SBF_i

The same approach can be applied for SBFC_m and SBFP_{cp}. However, ADUC_m and ADUP_{cp} must be determined from ADU_i.

2) Step 2: Calculation of ADUC_m and ADUP_{cp}

Regarding ADUC_m and ADUP_{cp}, the DDAE methodology does not propose a clear calculation step [1], [3]. and especially when the decoupled lead time DLT between two buffers is long (greater than a month).

We therefore propose to calculate ADUC_m at time T (and similarly ADUP_{cp}) by performing a bill-of-material explosion that considers decoupled lead times as indicated in the following formula:

$$ADUC^T = \sum_i \left(\sum_j ADU_i^{T+DLT_i} \cdot \alpha_{ij} \cdot BOM_{jm} \right)$$

Figure 6. illustrates this principle that can be integrated in each link between SBF_i, SBFC_m and SBFP_{cp} to compute ADUs adequately and to specify the relevant buffers' parameters.

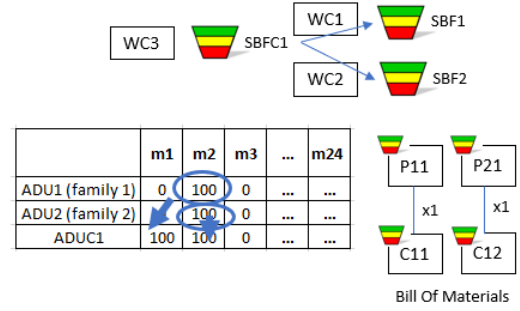


Figure 6. Buffer Profiles and Individual Part Properties of Stock Buffer [13].

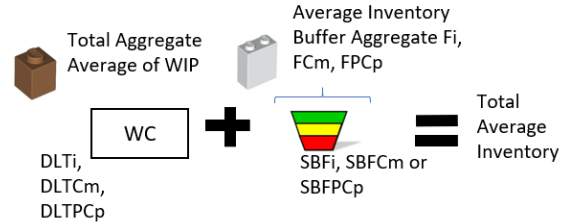


Figure 7. Total Average Inventory per month for calculating Working Capital.

3) Step 3: KPI to make a decision

In this step, the KPIs recommended by [3] are instantiated for financial perspective on one hand through working capital and contribution margin and for capability perspective on the other hand through load of critical resource.

a) Working Capital (WC)

Figure 7. shows the calculation of the overall average stock over a period studied with the following logic:

$$\text{Total Average Inventory} = \text{Total Average Inventory Buffer Family} + \text{Work in Progress}$$

For a stock buffer associated to a given product P_{ij}, he average stock is $\frac{GZ_{ij}}{2} + RZ_{ij}$, the in production is $DLT_{ij} \cdot ADU_{ij}$. We then propose the following formula at the aggregate product family level:

$$WC = \sum_i \left(\left(\frac{GZ_i}{2} + RZ_i + DLT_i \cdot ADU_i \right) \cdot PR_i \right) + \sum \left(\left(\frac{GZC}{2} + RZC + DLTC \cdot ADUC \right) \cdot PRC \right) + \sum_p \left(\left(\frac{GZPC_p}{2} + RZPC_p + DLTPC_p \cdot ADUPC_p \right) \cdot PRPC_p \right)$$

b) Contribution Margin (CM)

This KPI is used to make choices in case of saturation of the critical resource WOC_{crit} by prioritizing the ratio βF_i which has the largest contribution margin in money. This is an important criterion that allows the main services (Portfolio, Demand, Production, Purchasing...) participating in the AS&OP process to find the most profitable compromise. Therefore, we defined the parameter

θF_i as being the percentage of participation of each family F_i to the ADUCri corresponding to the critical resource. We finally propose the following formula principle at aggregate product family level:

$$CM = \sum_i (\theta F_i \cdot CM_{i..} \cdot ADUCrit_i)$$

c) Load of critical resource

As the maximum capacity of the critical resource $CapaWOCrit$ is known, it is easy to compare it with the load ADUCrit for each period studied throughout the AS&OP process. As for the contribution margin, we used appropriate ADU_i , $ADUC_m$ or $ADUC_p$ depending on $WOCrit$ position. This approach also makes it possible to check resources like operator, work-center or storage capacity.

IV. EXPERIMENT AND DISCUSSION

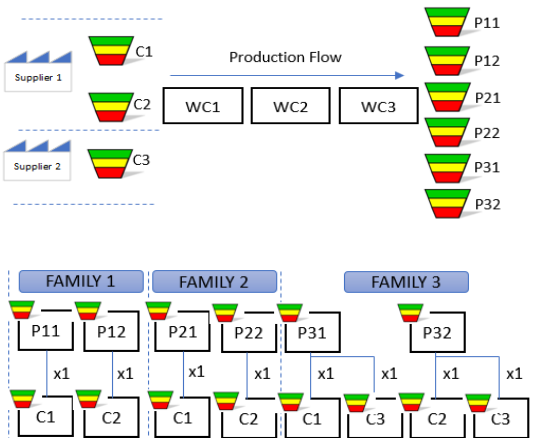
A. Illustrative Case Presentation

In order to validate our proposal, we consider a flow production workshop consisting of WOC_r work-centers with $r \{1,3\}$, a unique WH warehouse with products P_{ij} divided into Families F_i with $i \{1,3\}$ and $j \{1,2\}$ as well as components purchased C_k with $k \{1,3\}$. The critical resources are the following ones: WOC_r2 and WH. Each P_{ij} and C_k are individually managed by a stock buffer.

This illustrative case respects all the assumptions discussed in section III. Figure 8. describes the data set used for this experiment.

B. Implementation of Our Aggregate Model in Our Use Case

In the DDAE logic, rather than working on the basis of a single data set, including bias and uncertainties, the method suggests to always consider "roughly right" decision making process rather than "precisely wrong" decision making process [4]. Table II. Represents an anthology of opportunities and risks that a company must manage in an S&OP process. We will use them to create a scenario to test and validate our aggregate approach.



Family	Product	Sales volume	Saisonnality	Market	Price	contribution margin	DLT	Variability
F1	P11	High	non	Market 1	low	low	long > 1 month	low
	P12	High	yes	Market 1	low	low	long > 1 month	low
F2	P21	Low	no	Market 1	low	middle	middle < 1 month	middle
	P22	Low	no	Market 1	low	middle	middle < 1 month	middle
F3	P31	Sporadic	no	Market 2	high	high	middle < 1 month	high
	P32	Sporadic	no	Market 2	high	high	middle < 1 month	high

Supplier	Component	DLT	Variability	Buying Cost
Supplier 1	C1	short	low	low
	C2	short	low	low
Supplier 2	C3	long	High	high

Work Center	Critical Resource
WC1	no
WC2	yes
WC3	no
WH	yes

Demand		Ressource Planning			Finance		Operating Model				
Family	Product	WC1 (min/p)	WC2 (min/p)	WC3 (min/p)	Cube Utilization (m3)	Price (Dollars)	Margin Contribution (Dollars)	DLT	LTF	FV	MOQ
F1	P11	0,3	0,8	0	0,01	30	15	60	0,4	0,5	50
	P12	0,3	0,7	0	0,01	30	15	60	0,4	0,5	50
F2	P21	0,3	1,2	0,1	0,01	30	20	20	0,2	0,75	50
	P22	0,3	1,3	0,1	0,01	30	20	20	0,2	0,75	50
F3	P31	0,3	2	0,5	0,01	100	60	20	0,2	0,8	10
	P32	0,3	2,5	0,5	0,01	100	60	20	0,2	0,8	10

Supplier	Component	Cube Utilization (m3)	Price	DLT	FD	FV	MOQ	Capacity	min per day	m3
Supplier 1	C1	0,005	2	5	0,7	0,3	200	WC1	480	
	C2	0,005	3	5	0,7	0,3	200	WC2	480	
Supplier 2	C3	0,005	20	30	0,3	0,75	200	WC3	480	
								WH		250

Figure 8. Principal data of our theoretical data

TABLE II. EXAMPLE OF OPPORTUNITIES/RISKS ADRESSED DURING AS&OP

Notation	
OPPORTUNY	RISK
Expand to a new market	Consumer demand change, incorrect forecast, Lost Sales
New Product Introduction	Price Changes
Unique exceptional sale	Lack capacity (human, technology, process, business process, development problem...)
Change technology	Incorrect Inventory

TABLE III. EXAMPLE OF DECISION IN AS&OP INSPIRED BY [22]

	AREA DECISION	Example
OPERATION STRATEGY DECISION	CAPACITY Concerns how capacity and facilities in geneal should be configured	Adjust/Change Capacity (human or process)
	SUPPLY NETWORK Including Purchasing and Logistics Concerns how operations relate to the interconnected network of other operations"	"Make or Buy product Change supplier contract"
	PROCESS TECHNOLOGY Concerns to convert resources into finished goods or services	Integrate new technology
	ORGANIZATIONAL IMPROVEMENT Concerns to convert organization in an better one	Kaizen
MARKETING STRATEGY	INVENTORY MANAGEMENT Concerns how operations manage the management of production flows	Change operating model parameters
	PORTEFOLIO, NEW ACTIVITIES and DEMAND	"Change price Promotion

	<i>AREA DECISION</i>	<i>Example</i>
	Concerns how the market should be configured	activities accelerate the sale of a product"
<i>FINANCE STRATEGY</i>	FINANCE concerns the management of financial objectives	change / adapt the level of financial objectives to be achieved over a given period

Using the three-step methodology developed in section III, we obtain the aggregate model shown on figure 9.

The context for this AS&OP iteration is the following: an opportunity on market 2 concerning F3 is possible from the 10th month, with a 300% increase of volume. The decision-makers want to assess the potential impacts of such an increase and the associated decisions which can be made. The constraints are the following ones: not to exceed the maximum capacity of the critical resources WC2 and WH, not to exceed the working capital limit per month fixed by the finance department (1,600,000 \$), to realize profits greater than without this increase, improvements may be considered subject to not exceeding the limit set by the finance department.

TABLE IV. SYNTHESIS OF FOUR SCENARIOS TESTED VIA OUR MODEL ON OUR USE CASE

<i>Critical Resource WC2</i>	<i>Critical Resource WH</i>	<i>Working Capital</i>	<i>Margin Contribution</i>	<i>Test Decision</i>	
NOK between 13 to 25 hours necessary per month	OK	NOK exceeds the financial capacity 20,000 dollars per month	OK +4000 dollars per day	validation for this increase without others decisions	Scenario 1
OK with 90/100% utilization	OK Win 25% free capacity	OK 1,300,000 dollars per month	OK +4000 dollars per day	Increase capacity of WC2 with overtime of 25 hours by month Integrate an organizational improvement with DLT1 reduction of 10 days, give a new LTF1 to 0,4 to 0,3 for the family 1	Scenario 2
OK with 90/100% utilization	OK	NOK 1,650,000 dollars per month	OK + 6000 dollars per day	Increase 10 dollars the price of product family 1 Decrease 30% the sales forecasts of family 1 per month from the 10th month	Scenario 3
OK with 90/100% utilization	OK	NOK 1,650,000 dollars per month	OK + 6000 dollars per day	Scenario 3 with Risk decrease capacity 20% warehouse	Scenario 4

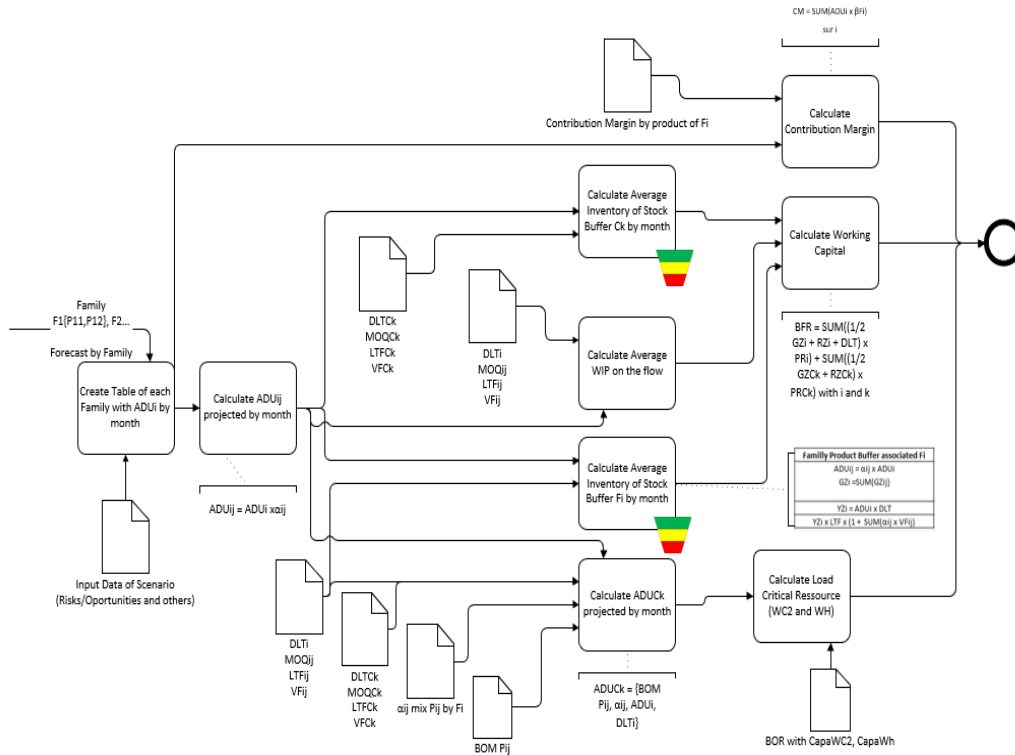


Figure 9. Specific Use Case Model Aggregate with the use of generic bricks of our aggregate model developed in section III

Four example scenarios have been successfully tested, mixing actions related to different possible decisions (Table IV.). With the use of a good level of information to the aggregate family product level, it is easy to build a scoreboard quickly interpretable by AS&OP stakeholders as shown on figure 10.

C. Discussions and Limit

Firstly, the proposed model and the previous illustrative case demonstrate that using an aggregate perspective to execute an AS&OP process is possible. The model suggested in this paper also gives a first concrete and

generic technical support to do this. However, this result has been done under specific conditions that must be discussed.

Secondly, concerning the assumptions made, it appears that some are questionable regarding real business cases. The assumption (A2) that consists in knowing α_{ij} might be difficult to validate. In some situation, gathering the volume estimate by family can be really difficult on a long-time range and would need a sensitivity analysis of the effects of estimation errors. The assumption (A3) considering that each stock is managed by a buffer is strong. The set of experience feedback presented during the various events organized for several years by the DDI [10] show i. taking into account time buffer and capacity buffer in addition to the stock buffer ii. rather hybrid solutions with some items managed through a DDOM, while others are managed through a MRP2. Our model does not yet integrate this hybrid situation. The assumption (A7) which consist in defining aggregates or families is sensitive. Actually, most of the companies which work with MRP2 and the associated S&OP process are normally used to do so. But, as indicated by [23], [24] often, the aggregates are simple aggregations of elementary end items data sets implying loss of information. This is typically the trap that consist in thinking in a “precisely wrong” perspective instead of reasoning in a “roughly right” dynamic[3]. Our solution which combines DDOM model with an aggregate planning production approach provides interesting alternative for practitioners.

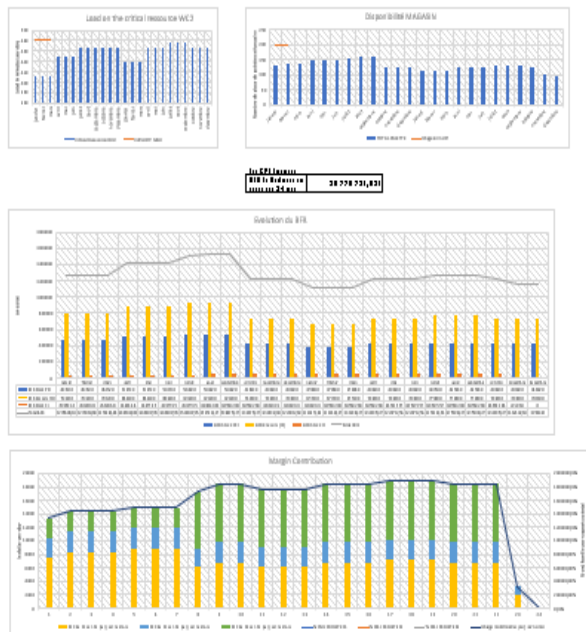


Figure 10. Simple Excel Scoreboard with relevant KPI inspired by DDAE on our model.

Thirdly, in the context of our illustrative case, we tested only few parameter changes. In a real context, the combinatory would be much larger and would imply more

variability and sensitivity in the decision-making process. In order not to miss out on a set of appropriate decisions, an automatic generation of several decision sets able to yield the most plausible results would be a real plus for supporting end users.

Fourthly, changes in the DDOM model, a tested scenario led us imagining an increase in demand on a family for a short time but with a high rate of demand variability (risk), as an exceptional opportunity to seize or not. This scenario involved modifying the forecast parameters of the DDOM at the same time as the increase in demand. This is an unusual approach in MRP2, linking strategic decisions at the S&OP level to parameterization decisions of the operating model in place. It is a specificity of the DDAE which multiplies the possibilities of actions. We sought to find out more about this on the basis of practitioner interview, the following decisions specific to the DDOM model may require the level of responsibility AS&OP: (i) change the strategic position of a buffer, (ii) the type of buffer (Time, Stock or Capacity), (iii) the calculation formula of ADU, (iv) horizon of detection of a spike, (v) the membership of an article to a profile of buffer but also to implement requests demand adjustment factor (DAF) or planning adjustment factor (PAF) of ADU in long period. Our model has challenged some of these elements, but not all. In particular, the following elements are not taken into account: time and capacity buffer, spike detection horizon, and will have to be taken into account to apply our approach to cover a wider spectrum of corporate reality.

V. CONCLUSION

In a context where the DDAE model is becoming more and more important to manage material flow in Supply Chains, we studied in this paper the following research question: *Is it possible to support an AS&OP process through a pure aggregate perspective? If yes, how this approach can be technically supported?* Our research work demonstrated that such an approach is possible and developed a concrete model able to support this in practice.

However, this work is its infancy and has still important limits, particularly regarding some assumptions made that are only partially validated by current real DDAE business cases.

Although this proposal has great potential for future DDAE implementations, it is also allowed identifying numerous rooms for future research among which:

Firstly, we have demonstrated that it is possible to achieve a holistic approach to support the AS&OP process, but it remains to demonstrate that it is an improvement over the direct use of the DDOM model and its simple data aggregation. A robust comparative study of these two approaches is needed to determine the best option for AS&OP.

Secondly, the experiment is quite poor at this stage and extensions seem to be mandatory. Notably, more complex

use cases included the following features would be interesting to study:

- Different types of buffer (Min/Max, Time Buffer, Capacity Buffer);
- Integrate peak demand variable as the DDAE manages it at the DDOM level;
- Environment with hybrid material management methods (KANBAN, CONWIP, DDOM, MRP2...)

Thirdly, our intention was to use aggregates to support the AS&OP process rather than an aggregated reference. Our model does this only partially. Seeking to eliminate steps using aggregated references is a way to go.

REFERENCES

- [1] C. Ptak et C. Smith, Demand Driven Material Requirements Planning (DDMRP), Version 3. Industrial Press, Incorporated, 2019.
- [2] R. Miclo, M. Luras, F. Fontanili, J. Lamothe, et S. A. Melnyk, « Demand Driven MRP: assessment of a new approach to materials management », *Int. J. Prod. Res.*, vol. 57, no 1, p. 166-181, janv. 2019.
- [3] D. Smith et C. Smith, Demand Driven Performance. McGraw Hill Professional, 2014.
- [4] G. R. Bitran et D. Tirupati, « Chapter 10 Hierarchical production planning », in *Handbooks in Operations Research and Management Science*, vol. 4, Elsevier, 1993, p. 523-568.
- [5] A. C. Hax et H. C. Meal, « Hierarchical Integration of Production Planning and Scheduling. », MASSACHUSETTS INST OF TECH CAMBRIDGE OPERATIONS RESEARCH CENTER, TR-88, sept. 1973.
- [6] H. ANDERSSON, S. AXSÄTER, et H. JÖNSSON, « Hierarchical material requirements planning », *Int. J. Prod. Res.*, vol. 19, no 1, p. 45-57, janv. 1981.
- [7] C. T. Maravelias et C. Sung, « Integration of production planning and scheduling: Overview, challenges and opportunities », *Comput. Chem. Eng.*, vol. 33, no 12, p. 1919-1930, déc. 2009, doi: 10.1016/j.compchemeng.2009.06.007.
- [8] M. Kremer, C. Schneeweiss, et M. Zimmermann, « On the validity of aggregate models in designing supply chain contracts », *Int. J. Prod. Econ.*, vol. 103, no 2, p. 656-666, oct. 2006, doi: 10.1016/j.ijpe.2005.12.007.
- [9] A. Cheraghalikhani, F. Khoshalhan, et H. Mokhtari, « Aggregate production planning: A literature review and future research directions », *Int. J. Ind. Eng. Comput.*, vol. 10, no 2, p. 309-330, 2019.
- [10] « Demand Driven Adaptive Enterprise (DDAE) », The Demand Driven Institute - World Leader in Demand Driven Education. Demanddriveninstitute.com, 2019. [Online]. Available: <https://www.demanddriveninstitute.com/demand-driven-adaptive-enterprise-m>. [Accessed: 11-Dec-2019].
- [11] G. Martin, P. Baptiste, J. Lamothe, R. Miclo, et M. Luras, « A process map for the demand driven adaptive enterprise model: towards an explicit cartography », 2018.
- [12] O. Wight, Manufacturing Resource Planning: MRP II: Unlocking America's Productivity Potential. John Wiley & Sons, 1995.
- [13] R. Miclo, Challenging the « Demand Driven MRP » Promises: a Discrete Event Simulation Approach. Ecole nationale des Mines d'Albi-Carmaux, 2016.
- [14] L. Ducrot et E. Ahmed, « Investigation of potential added value of DDMRP in planning under uncertainty at finite capacity », 2019.
- [15] L. Meinel, « DDMRP: presentation of a new solution of stock management and master production scheduling », juill. 2019.
- [16] A. P. V. Acosta, C. Mascle, et P. Baptiste, « Applicability of Demand-Driven MRP in a complex manufacturing environment », *Int. J. Prod. Res.*, vol. 0, no 0, p. 1-13, août 2019.
- [17] J.-B. Vidal, M. Luras, J. Lamothe, et R. Miclo, « Adaptive Sales & Operations Planning: Innovative Concept for Manufacturing Collaborative Decisions? », in *Collaborative Networks of Cognitive Systems*, 2018, p. 362-374.
- [18] A. M. Tavares Thomé, L. F. Scavarda, N. S. Fernandez, et A. J. Scavarda, « Sales and operations planning: A research synthesis », *Int. J. Prod. Econ.*, vol. 138, no 1, p. 1-13, juill. 2012.
- [19] N. Tuomikangas et R. Kaipia, « A coordination framework for sales and operations planning (S&OP): Synthesis from the literature », *Int. J. Prod. Econ.*, vol. 154, p. 243-262, août 2014.
- [20] J. Kristensen et P. Jonsson, « Context-based sales and operations planning (S&OP) research: A literature review and future agenda », *Int. J. Phys. Distrib. Logist. Manag.*, vol. 48, no 1, p. 19-46, janv. 2018.
- [21] P. Pittman et J. B. Atwater, APICS Dictionary, 15th Edition. APICS, 2016.
- [22] N. Slack et M. Lewis, Operations Strategy. Pearson Education, 2002.
- [23] R. C. Ling et W. E. Goddard, Orchestrating Success: Improve Control of the Business with Sales & Operations Planning. Wiley, 1989.
- [24] « CPIM - Certified in Production and Inventory Management | APICS », 2019, [Online]. Available sur: <http://www.apics.org/credentials-education/credentials/cpim>. [Accessed: 11-Dec-2019].