

APPLICATION OF MULTI-ITEMS INVENTORY MODELS IN COURSE MATERIALS MANAGEMENT: TOWARDS BUILDING A DECISION SUPPORT SYSTEM FOR THE NATIONAL OPEN UNIVERSITY OF NIGERIA.



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Received: September 21, 2022 **Accepted:** November 12, 2022

Abstract:

The classic multi-item inventory model with space, number of inventory and monetary constraints were applied in this research to solve the Course Materials inventory management problem of the National Open University of Nigeria (NOUN). The Kuhn-Tucker necessary and sufficient condition of determining optimum Economic Order Quantities (EOQ) which uses the Lagrange multiplier trial-and-error method was replaced in this work with a simple trimming method to speed up the process. Python programming was used to code the resultant model to give a flexible interface that will be easy to use by all users. Numerical illustrations were given for the three different models to validate the performance of the model on the test data. The trimming method used in this work was tested by comparing results obtained using the method with those obtained using the Lagrange trial-and-error method. When implemented, the proposed model will improve decision making in Course Material inventory management and reduce the difficulty encountered by staff in utilization of stored data for basic scientific analysis using manual computational processes.

Keywords:

Course Materials, Decision Support System, Economic Order Quantity, Multi-Item Inventory Models, Inventory Constraint, Python Programming, Space Constraint

Introduction

Managing the complex processes involved in production and Inventory control using human intuitive judgment and manual processes is undoubtedly daunting with too many associated challenges. More so, these manual processes most times produce unreliable outputs which lead to wastages, extra costs and other disadvantages compared with automated systems. The National Open University of Nigeria is faced with the challenge of managing its over one thousand eight hundred different Course Materials (CM) which they are obligated to print and distribute to students every semester after registration. NOUN offers wide range of programmes across its eight faculties with over Five Hundred Thousand students. These students are spread across over one hundred Study Centres nationwide. Every Semester, there is need to distribute over One Thousand Eight Hundred distinctive printed Course materials to these students at their different locations to enable them study and prepare for examination.

A study of the system of Course Materials production planning and inventory management in the National Open University of Nigeria shows that a lot of problem-solving and decision-making processes are manually carried out. This has effect on the overall productivity and efficiency of the resultant system as evident in the non-availability of some Course Materials at the Study Centres when demanded by students. There is need therefore, to develop a comprehensive decision support system (DSS) for problem solving and decision making in the entire processes of production and inventory management of Course Materials in NOUN. Such decision support system must consists of relevant models and tools for forecasting students population, determining demand for each Course Material, managing production and inventory of the Course Materials. Adoga et al. (2022) discussed the demand forecast module and the Economic production module of the proposed DSS, in this paper, we present the inventory management module of the decision support system which will be used for the Course Materials inventory management.

Inventory management problems exist in many organizations that rely on inventory items for day to day running of the organizations. Such organizations are usually faced with the problem of how much items to order so as to service their demands for a certain period of time and when to place such orders for each inventory item in order not to go out of stock. Inventory Management is important for the successful operations of most organizations due to the amount of money inventory represents. Efficient inventory and production planning hinges on the right decision being made at the right time as each wrong decision may cost the firm very heavily in terms of money, labour and other resources (Russel & Taylor, 1995). To avoid wasteful spending on inventory, a firm must establish an inventory policy that specifies when an order for additional items should be placed and how many items should be ordered at each time. In this regard, different types of inventory models have been developed to provide effective directional guidance for inventory management (Masood, 1998). This research therefore presents three specialized inventory models to help NOUN Decision Makers determine the optimum order quantities and when to place orders for each Course Material to ensure all year availability of the Materials.

The Economic Order Quantity model is generally used to determine an optimal order size by balancing the setup cost and inventory holding cost (Jonrinaldi et al., 2018). Kasthuri et al. (2011) developed a multi-item inventory model with constraints of storage space, number of orders and production cost in both crisp and fuzzy environment. The Kuhn-Tucker conditions method was employed in solving the model. The developed model can be modified to accommodate constraints such space, inventory level and budget. Rohatgi & Agrawal (2016) applied multi-item inventory models in the area of pharmacy inventory

management. The problem was derived as nonlinear programming problem under limited budget and number of orders in a fixed period of time constraints. The results proved effective in the computation of EOQ, Optimum cost and optimum number of orders for a list of pharmaceutical inventory items with different yearly demand. Miranda et al. (2015) presented an alternative approach for the calculation of the multi-item EOQ in presence of space restrictions. The model introduced an exit condition to stop the algorithm quickly when there is the certainty that the optimal solution has been found.

Hariga (2000) developed an efficient iterative procedure to determine the near optimal lot size, reorder point and setup time. The reduction in setup time proposed by the model is efficient in reduction of the total inventory costs and results in a smaller lot size and a shorter lead time. Ouyang et al. (1999) investigated the impact of ordering cost reduction on the modified continuous review inventory systems involving variable lead time with a mixture of backorders and lost sales. An algorithm procedure to find the optimal order quantity, ordering cost, reorder point and lead time simultaneously was finally developed. Lin & Hou (2005) proposed an inventory system with random yield in which both the setup cost and yield variability can be reduced through capital investment. The paper discussed a procedure to find the optimal order quantity, reorder point, optimal setup cost and yield standard deviation. Results showed that significant cost savings can be achieved by adopting capital investments.

Zhang & Wang (2011) explored the structural properties of deteriorated multi-item EOQ model and proposed an algorithm for solving the optimal solution by proving that the studied problem is a special convex separable nonlinear knapsack problem. The result is also applicable to solving the classical constrained multi-item EOQ problem with suitable modification. Ophokenshi et al. (2019) proposed a mathematical model of an inventory system with timedependent, three-parameter Weibull deterioration and a stochastic type demand in the form of a negative exponential distribution. The optimal inventory policy for the proposed model was derived and the necessary and sufficient condition for the optimal policy was also established. The objective of Kotb & Fergany (2011) was to derive the analytical solution of the EOQ model of multiple items with both demand-dependent unit cost and lead time using geometric programming approach. The optimal order quantity, the demand rate and the lead time were determined and used to obtain the optimal total cost.

Yen et al. (2019) presented a model that considers the holding costs of raw materials under conditions of two-level trade credit and limited storage capacity. Four theorems were developed in the work to characterize the optimal solutions according to a cost minimization strategy. Ukil & Uddin (2016) developed a time dependent inventory model on the basis of constant production rate and market demands which are exponentially decreasing. The model developed an algorithm to determine the optimal demand, optimal order interval, optimal time cycle and the optimum total cost.

Skowron & Styczen (2016) considered a complex autonomous inventory coupled system that can take the form of a network of chemical or biochemical reactors for example and where the inventory interactions perform the recycling of by-products between the subsystems. A multifrequency second-order test for complex multi-periodic systems including the boundary optimal steady-state process and an arbitrary large number of harmonics used to verify its improvement by the multi-periodic operation was also proposed.

Sharma (2005) discussed extensively the Lagrange method of solving Multi-Item Deterministic Inventory Models (The classical EOQ with Constraints models). The method of Lagrange's multiplier was used to minimize the cost equations considering limitations of storage, finance and inventories in separate cases. The systematic trial and error method was used to find λ^* (the optimum value of λ), by trying successive value of λ , the values of λ^* should result in simultaneous values of q_i^* satisfying the given constraint equations.

Materials and Methods

Here we present the steps followed in applying the classic multi-items inventory models to the data of NOUN course material inventory. The classic multi-item inventory models with space, monetary and inventory limitations discussed in Sharma (2005) were adopted with modifications in the slow trial-and-error method which were replaced with faster trimming methods in this work.

The Classical EOQ Model is constructed based on the conditions that goods arrive the same day they are ordered and no shortages allowed. Figure 1 show the classic inventory cycle. The cycle begins at time 0 with an arriving order. The amount of the order is the lot size, q. The lot is delivered all at one time causing the inventory to shoot from 0 to Q instantaneously. Materials are withdrawn from inventory at a constant demand rate, R. After the inventory is depleted, the time for another order of size Q arrives, and the cycle repeats (Taha, 2007). Figure 1 shows the cycle of a single item but there are many real life situations where multiple items are required to run an organization. These multiple items most times compete for space, monetary and other limited resources of the organization making the inventory problem more complex to deal with. In such situations, the EOQ models are first applied without constraints. Then the results are subjected to constraints depending on the nature of the problem.

Sharma (2005) gave the unconstraint equation for the order quantity q_i as:

$$q_i = \sqrt{\frac{2C_3^{(i)}R_i}{C_1^{(i)}}}$$
, $i = 1, 2, ..., n$. (1)

While the equation for the limitation on floor space (Storage space) was given as:

$$q_i^* = \sqrt{\left(\frac{2C_3^{(i)}R_i}{C_1^{(i)} + 2\lambda^* a_i}\right)}, i = 1, 2, ..., n.$$
 (2)

Subject to
$$\sum_{i=1}^{n} a_i q_i^* \le A$$
 (3)

The equations for limitation on investment, was given as:

$$q_i^* = \sqrt{\left(\frac{2C_3^{(i)}R_i}{C_1^{(i)} + 2\lambda^* C_4^{(i)}}\right)} \tag{4}$$

Subject to $\sum_{i=1}^{n} C_4^{(i)} q_i^* \le M$ (5)

Similarly, the equation for limitation of Inventories was

$$q_i^* = \sqrt{\frac{2C_3^{(i)}R_i}{C_1^{(i)}+\lambda^*}}, i = 1,2....,n.$$
 (6)

Subject to
$$\sum_{i=1}^{n} q_i^* \le N$$
 (7)
Finally, P, d and C(t) are given by:

$$P = \frac{R_{i}}{q_{i}^{*}}$$

$$d = \frac{365}{P}$$

$$C(t) = \frac{R_{i}}{q_{i}^{*}} C_{3}$$
(8)
(9)

$$d = \frac{365}{P} \tag{9}$$

$$C(t) = \frac{R_i}{\sigma_i^*} C_3 \tag{10}$$

Notations and Assumptions of the Models

n = number of items

 R_i = demand rate for the i^{th} item (i = 1, 2, ..., n).

 $C_1^{(i)}$ = holding (or carrying) cost per unit of the quantity of ith item.

 $C_3^{(i)}$ = Ordering cost per unit order for the ith item. $C_4^{(i)}$ = unit price of the ith item

C(q) = Total inventory cost

C(t) = Total annual ordering cost

 q_i = the amount of quantity ordered of the i^{th} item. q_i^* = Economic Order Quantity (EOQ) of the ith item.

 λ = Lagrange Multiplier

P = number of Orders per year

d = Number of days' supply per optimum order

T = trimming constant

A = the maximum storage area (in sq. meter) available for the n items.

 $a_i\!=\!Storage$ area required per unit of the i^{th} item

M = The Amount of money available for investment on inventory at a particular time.

N = Maximum number of stock at a given time. The models have the same assumptions with the classic inventory models as follows:

- The demand is deterministic and supply i. is instantaneous.
- ii. Shortages are not allowed.
- iii. The order of each item is not dependent on the other.

Applying the inventory model with limitation in space

In this model, the inventory system includes n > 1 items which are competing for a limited storage space. An interaction between the different items occurring due to this limitation is considered as a constraint. The classical multiitem inventory model uses the Kuhn-Tucker necessary and sufficient condition which uses the Lagrangian multiplier and trial-and-error methods to find the constrained optimum order quantities of each item from equation 2 and 3. The trial-and error method is a slow algorithm considering the

number of iterations that may occur before arriving at a solution. To fasten the process, this study determined q_i^* for each item using the steps below:

- Determine the unconstrained value of q_i using i. equation 1
- ii.
- Compute the value of $\sum_{i=1}^n a_i q_i$ If the value of $\sum_{i=1}^n a_i q_i \leq A$ stop and set the iii. obtained values q_i to q_i^* (optimum values)
- If $\sum_{i=1}^{n} a_i q_i > A$, we compute a trimming constant (T) given by $T = \frac{A}{\sum_{i=1}^{n} a_i q_i}$ (11) iv.
- q_i^* (optimum values) = Tq_i for i = 1, 2, ..., n. v.

vi. Check if
$$\sum_{i=1}^{n} Tq_i^* \le A$$

Applying the inventory Model with limitation in investment (monetary constraint)

In this case we consider an upper monetary limit M (in Naira) on the amount available for investment on the entire inventory items. To achieve the objective in equation 5, we apply our trimming method to obtain the optimum value of q_i^* after determining the unconstrained value of q_i using equation 1. The steps involved in executing the model are:

- ii.
- The steps involved in executing the model are: Compute the value of $\sum_{i=1}^{n} C_4^{(i)} q_i$ If the value of $\sum_{i=1}^{n} C_4^{(i)} q_i \leq M$ stop and set the obtained values q_i to q_i^* (optimum values)

 If $\sum_{i=1}^{n} C_4^{(i)} q_i > M$, we compute a trimming constant (T) given by $T = \frac{M}{\sum_{i=1}^{n} C_4^{(i)} q_i}$ (13) q_i^* (optimum values) = Tq_i for i = 1, 2, ..., n. Check if $\sum_{i=1}^{n} C_4^{(i)} q_i^* \leq M$ The inventory Model with limitation in iii.
- iv.

Applying the inventory Model with limitation in Inventories items

In this case we intend to limit the number of all units of the stocks to N,

i.e
$$\sum_{i=1}^{n} q_i \le N$$
, i = 1,2..., n.

Again we apply our trimming method to obtain the optimum value of q_i^* after determining the unconstrained value of q_i using equation 1. The steps involved in executing the model are:

- i.
- Compute the value of $\sum_{i=1}^{n} q_i$ If the value of $\sum_{i=1}^{n} q_i \leq N$ stop and set the obtained values q_i to q_i^* (optimum values) ii.
- If $\sum_{i=1}^{n} q_i > N$, we compute a trimming constant (T) given by $T = \frac{N}{\sum_{i=1}^{n} q_i}$ (14) q_i^* (optimum values) = Tq_i for i = 1, 2, ..., n. iii.
- iv.
- Check if $\sum_{i=1}^{n} q_i^* \leq N$

A flowchart of the model selection process is shown in figure 1.

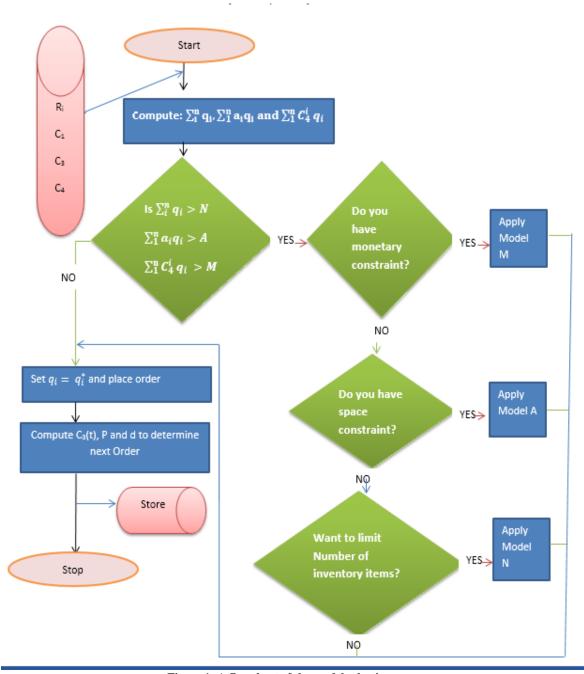


Figure 1: A flowchart of the model selection process

System Implementation

Jupyter notebook Integrated Development Environment (IDE) was used to develop the DSS prototype using python programming language. The inventory module of the Decision Support System was tested to determine the performance and accuracy of the developed system.

The inputs in table 2-4 were used to illustrate the performance of the three multi-items inventory models discussed in this research. The values of A, M and N in equations 11, 13 and 14 were given as $12m^2$, $\frac{1}{N}$ 2, 500,000

and 1800 units respectively. The Trimming factor for each model was computed from the respective constraint equations as follows:

T =
$$\frac{A}{\sum_{i=1}^{n} a_i q_i}$$
 = 0.5737659877159055, $T = \frac{M}{\sum_{i=1}^{n} C_4^{(i)} q_i}$ = 0.5341247129251392, and , $T = \frac{N}{\sum_{i=1}^{n} q_i}$ = 0.860648981573858

for the space, monetary and inventory constraint models respectively. The performance of the trimming method used in this work was evaluated relative to the Lagrange multiplier parameter method found in literature. An example of each constraint model solved using trial and error Lagrange multiplier method (Sharma, 2005) was reworked using the inbuilt trimming factor of our DSS. Results of both methods were compared as shown in table 1.

Results and Discussion

Tables 2-4 also show the outputs of the inventory module of the developed DSS prototype. The inventory module has three inbuilt inventory models; the models with space, monetary and inventory constraints. The values of EOQ (q_i^*) of each item shown in tables 2-4 represents the optimum number of items required per order so as to meet the projected yearly demand without stock out . (P) Represents the optimum number of orders per year that will satisfy the total yearly demand while C(q) represents the cost of ordering the EOQ of each item per order made. These results provide answers to basic question such as when to Order, how much to Order and the cost of each Order. For every Order, the EOQ of each item is ordered and received. The results are in line with the inventory cycle discussed by Taha (2007) and will serve as a basis for formation of an inventory

policy for the institution. Table 1 show the results obtained when the trimming method used in this research was compared with the trial and error Lagrange multiplier method presented in Sharma (2005). The EOO obtained from our DSS inventory module and that obtained by Sharma (2005) from the same inputs were compared. The variations from the two methods were relatively small and negligible. Our proposed trimming method is therefore recommended for its speed over the slow trial and error method which in some cases took up to seven trials before the final result. The interface of the developed DSS prototype allows users to select an input file for an inventory model and then provide tools for execution of the chosen model. Upon execution, the model outputs consisting of the economic order quantities (EOQ) of each course material, optimum number of orders per year and the total ordering cost are displayed in the model output window. The results of this module will help Decision Makers to decide on when to place orders and how much quantity of each item to order at a given time.

Table 1: Validation of the Space Model

Item code	Annual	Ordering	Holding	Storage	q_i	EOQ	EOQ	Positive
(Inventory item)	Forecast Demand (Ri)	Cost per Unit item (C3)	Cost per unit item (C1)	area reserve for the ith item (ai)		$(qi^* = Tq_i)$	computed from Lagrange Multiplier method	difference
1	20	100	30	1	11.54701	5.5	6.7	1.2
2	40	50	10	1	20	9.5	7.6	1.9
3	30	150	20	1	21.2132	10.0	10.6	0.6
Grand Total					52.76021	25	24.9	0.1

Table 2: Inputs and Outputs of the multi-item EOQ model with space limitation

Course	Annual	Ordering	Holding	Storage	Unconstraint Order	EOQ	Number	C3(t)
code (items)	Forecast Demand	Cost per Unit	Cost per unit item	area reserve for	Quantity $q_i =$	(qi* = Ta)	of Orders	
(Itellis)	(Ri)	item	(C1)	the ith item	$2C_3^{(i)}R_i$	Tq_i)	per year (P)	
	(14)	(C3)	(01)	(ai)	$\sqrt{\frac{\frac{2C_3-R_l}{C_1^{(i)}}}{C_1^{(i)}}}$		(1)	
CIT101	1506	400	200	0.01	77.61443	44.532	33.81798	13527.19
CIT102	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
CSS111	1342	400	200	0.01	73.26664	42.037	31.92357	12769.43
CSS112	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
CSS121	1342	400	200	0.01	73.26664	42.037	31.92357	12769.43
CSS132	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
CSS133	1342	400	200	0.01	73.26664	42.037	31.92357	12769.43
CSS134	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
CSS136	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
CSS152	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
ECE110	273	400	200	0.01	33.04542	18.960	14.39848	5759.39
ECE112	273	400	200	0.01	33.04542	18.960	14.39848	5759.39
ECE113	271	400	200	0.01	32.92416	18.890	14.34564	5738.255
ECE120	273	400	200	0.01	33.04542	18.960	14.39848	5759.39
ECE121	271	400	200	0.01	32.92416	18.890	14.34564	5738.255
ECE123	271	400	200	0.01	32.92416	18.890	14.34564	5738.255
ECO121	1342	400	200	0.01	73.26664	42.037	31.92357	12769.43
EDU111	435	400	200	0.01	41.71331	23.933	18.17523	7270.091
EDU112	441	400	200	0.01	42	24.098	18.30014	7320.057
EDU114	441	400	200	0.01	42	24.098	18.30014	7320.057
ENG111	164	400	200	0.01	25.6125	14.695	11.15982	4463.927
ENG113	164	400	200	0.01	25.6125	14.695	11.15982	4463.927
ENG114	168	400	200	0.01	25.92296	14.873	11.29509	4518.038
ENG121	164	400	200	0.01	25.6125	14.695	11.15982	4463.927
ENG122	168	400	200	0.01	25.92296	14.873	11.29509	4518.038
ENG141	164	400	200	0.01	25.6125	14.695	11.15982	4463.927
ENG162	168	400	200	0.01	25.92296	14.873	11.29509	4518.038
ENG172	168	400	200	0.01	25.92296	14.873	11.29509	4518.038
GST101	1777	400	200	0.01	84.30896	48.373	36.7349	14693.96
GST102	1738	400	200	0.01	83.37865	47.839	36.32956	14531.82
GST105	1777	400	200	0.01	84.30896	48.373	36.7349	14693.96
GST107	1777	400	200	0.01	84.30896	48.373	36.7349	14693.96
PCR111	1342	400	200	0.01	73.26664	42.037	31.92357	12769.43
PCR114	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
PED112	273	400	200	0.01	33.04542	18.960	14.39848	5759.39
PED122	271	400	200	0.01	32.92416	18.890	14.34564	5738.255
PED130	273	400	200	0.01	33.04542	18.960	14.39848	5759.39
PED144	271	400	200	0.01	32.92416	18.890	14.34564	5738.255
POL111	1342	400	200	0.01	73.26664	42.037	31.92357	12769.43
POL126	1297	400	200	0.01	72.02777	41.327	31.38378	12553.51
Grand Total	1271	100	200	0.01	2091.445	1200	NA	364511.8

Table 3: Inputs and Outputs of the multi-item EOQ model with monetary limitation

Course code (Inventory item)	Annual Forecast Demand (Ri)	Ordering Cost per Unit item (C3)	Holding Cost per unit item (C1)	unit price of the ith item (C4)	$q_{i} = \sqrt{\frac{2C_{3}^{(i)}R_{i}}{C_{1}^{(i)}}}$	EOQ (qi* = Tq_i)	Number of Orders per year (P)	C(t)
CIT101	1506	400	200	2000	77.61443	41.45579	36.32786	14531.14
CIT102	1297	400	200	2000	72.02777	38.47181	33.71299	13485.2
CSS111	1342	400	200	2500	73.26664	39.13352	34.29285	13717.14
CSS112	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
CSS121	1342	400	200	2500	73.26664	39.13352	34.29285	13717.14
CSS132	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
CSS133	1342	400	200	2500	73.26664	39.13352	34.29285	13717.14
CSS134	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
CSS136	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
CSS152	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
ECE110	273	400	200	2000	33.04542	17.65038	15.46709	6186.837
ECE112	273	400	200	2000	33.04542	17.65038	15.46709	6186.837
ECE113	271	400	200	2000	32.92416	17.5856	15.41033	6164.133
ECE120	273	400	200	2000	33.04542	17.65038	15.46709	6186.837
ECE121	271	400	200	2000	32.92416	17.5856	15.41033	6164.133
ECE123	271	400	200	2000	32.92416	17.5856	15.41033	6164.133
ECO121	1342	400	200	2500	73.26664	39.13352	34.29285	13717.14
EDU111	435	400	200	2000	41.71331	22.28011	19.52414	7809.657
EDU112	441	400	200	2000	42	22.43324	19.65833	7863.332
EDU114	441	400	200	2000	42	22.43324	19.65833	7863.332
ENG111	164	400	200	2000	25.6125	13.68027	11.98807	4795.228
ENG113	164	400	200	2000	25.6125	13.68027	11.98807	4795.228
ENG114	168	400	200	2000	25.92296	13.8461	12.13338	4853.354
ENG121	164	400	200	2000	25.6125	13.68027	11.98807	4795.228
ENG122	168	400	200	2000	25.92296	13.8461	12.13338	4853.354
ENG141	164	400	200	2500	25.6125	13.68027	11.98807	4795.228
ENG162	168	400	200	2000	25.92296	13.8461	12.13338	4853.354
ENG172	168	400	200	2500	25.92296	13.8461	12.13338	4853.354
GST101	1777	400	200	2000	84.30896	45.0315	39.46127	15784.51
GST102	1738	400	200	2000	83.37865	44.5346	39.02584	15610.33
GST105	1777	400	200	2000	84.30896	45.0315	39.46127	15784.51
GST107	1777	400	200	2000	84.30896	45.0315	39.46127	15784.51
PCR111	1342	400	200	2500	73.26664	39.13352	34.29285	13717.14
PCR114	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
PED112	273	400	200	2000	33.04542	17.65038	15.46709	6186.837
PED122	271	400	200	2000	32.92416	17.5856	15.41033	6164.133
PED130	273	400	200	2000	33.04542	17.65038	15.46709	6186.837
PED144	271	400	200	2000	32.92416	17.5856	15.41033	6164.133
POL111	1342	400	200	2500	73.26664	39.13352	34.29285	13717.14
POL126	1297	400	200	2500	72.02777	38.47181	33.71299	13485.2
Grand Total					2091.445	1117.092	NA	391564.9

Table 4: Inputs and Outputs of the multi-item EOQ model with Inventory constraint

Course code	Annual Forecast	Ordering Cost per Unit item (C3)	Holding Cost per	Unconstraint Order Quantity	EOQ (qi* =	Number of Orders per	C(t)
(Inventory	Demand	, ,	unit item		Tq_i)	year (P)	
item)	(Ri)		(C1)	$(q_i = \sqrt{\frac{2C_3^{(i)}R_i}{C_1^{(i)}}})$			
CIT101	1506	400	200	77.61443	66.79878	22.54532	9018.129
CIT102	1297	400	200	72.02777	61.99063	20.92252	8369.007
CSS111	1342	400	200	73.26664	63.05686	21.28238	8512.952
CSS112	1297	400	200	72.02777	61.99063	20.92252	8369.007
CSS121	1342	400	200	73.26664	63.05686	21.28238	8512.952
CSS132	1297	400	200	72.02777	61.99063	20.92252	8369.007
CSS133	1342	400	200	73.26664	63.05686	21.28238	8512.952
CSS134	1297	400	200	72.02777	61.99063	20.92252	8369.007
CSS136	1297	400	200	72.02777	61.99063	20.92252	8369.007
CSS152	1297	400	200	72.02777	61.99063	20.92252	8369.007
ECE110	273	400	200	33.04542	28.44051	9.598984	3839.594
ECE112	273	400	200	33.04542	28.44051	9.598984	3839.594
ECE113	271	400	200	32.92416	28.33614	9.563758	3825.503
ECE120	273	400	200	33.04542	28.44051	9.598984	3839.594
ECE121	271	400	200	32.92416	28.33614	9.563758	3825.503
ECE123	271	400	200	32.92416	28.33614	9.563758	3825.503
ECO121	1342	400	200	73.26664	63.05686	21.28238	8512.952
EDU111	435	400	200	41.71331	35.90052	12.11682	4846.727
EDU112	441	400	200	42	36.14726	12.2001	4880.038
EDU114	441	400	200	42	36.14726	12.2001	4880.038
ENG111	164	400	200	25.6125	22.04337	7.439879	2975.952
ENG113	164	400	200	25.6125	22.04337	7.439879	2975.952
ENG114	168	400	200	25.92296	22.31057	7.530063	3012.025
ENG121	164	400	200	25.6125	22.04337	7.439879	2975.952
ENG122	168	400	200	25.92296	22.31057	7.530063	3012.025
ENG141	164	400	200	25.6125	22.04337	7.439879	2975.952
ENG162	168	400	200	25.92296	22.31057	7.530063	3012.025
ENG172	168	400	200	25.92296	22.31057	7.530063	3012.025
GST101	1777	400	200	84.30896	72.56042	24.48994	9795.975
GST101	1738	400	200	83.37865	71.75975	24.2197	9687.882
GST10 2 GST105	1777	400	200	84.30896	72.56042	24.48994	9795.975
GST107	1777	400	200	84.30896	72.56042	24.48994	9795.975
PCR111	1342	400	200	73.26664	63.05686	21.28238	8512.952
PCR114	1297	400	200	72.02777	61.99063	20.92252	8369.007
PED112	273	400	200	33.04542	28.44051	9.598984	3839.594
PED122	271	400	200	32.92416	28.33614	9.563758	3825.503
PED130	273	400	200	33.04542	28.44051	9.598984	3839.594
PED144	271	400	200	32.92416	28.33614	9.563758	3825.503
POL111	1342	400	200	73.26664	63.05686	21.28238	8512.952
POL126	1297	400	200	72.02777	61.99063	20.92252	8369.007
Grand Total	12/1	.00	200	2091.445	1800	NA	243007.9

Conclusion

The classic multi-item inventory model with space, number of inventory and monetary constraints were applied in this research to solve the Course Materials inventory management problem of NOUN. The rather slow process of determining optimum Economic Order Quantities using the Lagrange multiplier trial-and-error method was replaced in this work with a trimming constant. The resultant model was

coded using python programming for easy use by concerned staff of NOUN. Integrating a wide range of inventory models in a single Decision Support System provides the decision maker with a wide range of alternatives thereby enhancing the overall decision making process.

Effective modeling of the Production and Inventory Management system of Course Materials in the National Open University of Nigeria will solve the problem of nonavailability of these important learning materials when demand is made for them by students. Also automating the processes involved in the production and management of these materials will enable staff and Managers to carry out their duties with minimal stress and technical requirement. Application of the models discussed in this work will significantly increase productivity, efficiency and give management enough time to budget and plan ahead towards meeting the demands for course materials and other services in a forecast year. Finally, the results of this research will breach significantly the dearth of knowledge in the area of application of model-driven Decision Support Systems in problem solving in Nigerian institutions.

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