A resilient approach to modelling the supply and demand of platelets in the United Kingdom Blood supply chain

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A resilient approach to modelling the supply and demand of platelets in the United Kingdom Blood supply chain

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A resilient approach to modelling the supply and demand of platelets in the United Kingdom Blood supply chain

For Peer Review Only

Abstract

The short shelf-life of platelets together with their stochastic demand and variable supply creates problematic decision making for inventory managers who seek to maintain optimal stock levels. This paper develops a novel simulation model that acts as a decision support tool for identifying optimal stock levels that would guarantee the availability of stock and, eliminate wastages and outages within the platelet supply chain. A multi-method approach is delineated upon that combines agent-based and discrete event modelling built on Anylogic software using data collected from typical UK Stock Holding Units (SHUs). The model simulates the demand and supply of 24 different platelet types broken down into their ABO and rhesus groupings and recommends the appropriate stock level at SHUs that will reduce expiries by 78% while simultaneously guaranteeing stock availability. This research not only represents the first attempt to fully understand the UK platelet supply chain but also presents an advanced simulation model that adopts a ‘pull’ system of platelet inventory management. Further, the concept of resiliency is integrated into the design of the solution methodology by testing the model behaviour in unexpected disruptions in the supply chain during an emergency case study.

Keywords: *Blood supply chain, platelets, simulation, inventory management, optimisation.*

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3 **1.0. Introduction**
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5 The supply of blood products is irregular (due to the uncertainty of donor willingness) and its
6 demand stochastic (Beliën & Forcé, 2012) making it difficult to accurately forecast
7 production and schedule blood donations (Samani et al., 2020). Therefore, inventory
8 management seeks a trade-off between shortages and wastages to guarantee the availability of
9 blood products while simultaneously minimizing expiries (Stanger, et al., 2012). Although
10 more than a hundred different products can be extracted from blood, platelets, red blood cells
11 and plasma are the most important components (Katsaliaki & Brailsford, 2017). Platelets are
12 particularly significant because they have multiple uses in chemotherapy, bone marrow
13 transplants, treatment of coronary artery diseases and more recently, researchers have
14 discovered its application in developing a Coronavirus (COVID-19) vaccine (Chang et al.,
15 2020).

16 One challenge for inventory managers, however, is that platelets have the highest rate of
17 perishability- with a short five to seven days storage shelf-life (Zahiri, et al., 2018). Indeed, in
18 the USA and Western Europe, approximately 20% of platelets collected result in expiries
19 (Rajendran & Ravindran, 2020). Apart from wastages, unpredictable challenges caused by
20 unplanned surgeries, natural disasters, equipment breakdown, war and labour strikes lead to
21 significant disruptions and losses within the blood supply chain and consequently, unsatisfied
22 demands (Haeri et al., 2020). It is crucial therefore to public health, that policies are adopted
23 that assimilate the concept of ‘resilience’ (i.e. an innate ability to withstand large-scale
24 disruptions) into planning and inventory management to remain reliable despite unexpected
25 challenges (Rajendran & Srinivas, 2020).

26 To resolve this aforementioned problematic public health conundrum and achieve
27 equilibrium between blood products supply and demand, this study focuses on platelet
28 inventory management and seeks to answer the following inductive questions: 1) What is the
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3 optimal level of platelet inventory that will meet demand and reduce shortages? 2) What is
4 the impact of relying on hospital demand to make decisions on inventory levels at Stock
5 Holding Units (SHUs) vis-à-vis depending on the supply side? 3) How can resilient inventory
6 management decisions be made in the face of an emergency?
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11 Although several authors have attempted to provide a solution to the platelet inventory
12 management problem, Beliën & Forcé (2012) report upon a discernable lack of fully scalable
13 models (both mathematical and simulation) that provide clear demonstrable evidence of
14 performance improvement in practice. Most innovative academic solutions are limited to
15 theoretical contexts only (Stanger, et al., 2012a) and research into blood inventory
16 management remains scant (Rajendran & Srinivas, 2020). Within the specific area of platelet
17 inventory management research, consideration of the different platelet categories, together
18 with their ABO and rhesus groupings remains unexplored. Studies predominantly devote
19 their analysis to one type of platelet product (Haeri et al., 2020), which is mostly apheresis
20 platelets. Furthermore, integrating the concept of resiliency into the design of a solution
21 methodology for platelet inventory management has not yet been fully addressed. Haeri et al
22 (2020) conducted perhaps the closest match to this present study and considered different
23 resiliency measures in the blood supply chain. However, their study (*ibid*) focuses on the
24 design of the blood supply chain network. Consequently, this current research utilizes data
25 collected from a real-life SHU in England to develop a pragmatic decision support tool that
26 can be implemented in generating optimal stock levels. In addition to being the first attempt
27 to fully understand the UK platelet supply chain, this novel study is also the first to propose a
28 multi-product simulation model that advances previous simulation models viz.: 1) It analyses
29 24 different types of platelet categories broken down into apheresis, pooled and paediatric
30 platelets, the different ABO categories and their rhesus groupings. While previous
31 researchers have developed bespoke solutions for the blood supply chain, their focus has
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mainly been on red blood cells and apheresis platelets; 2) The simulation model was designed to operate based on a ‘pull’ system meaning that the stock levels at the SHU are influenced by hospital demands. The pull system facilitates elimination of wastage and reduction of cost (Lu & Zheng, 2009) since the model’s outputs inform production planning; 3) Substitution of platelet products is considered; 4) The model’s resilience was tested in an emergency case study of a real-life SHU.

The rest of the paper is organised as follows: *section 2.0. The platelet supply chain: a literature review* synthesises relevant literature on platelet supply chain; *section 4.0. Materials and methods* presents the case study; *section 3.0. Case study description* delineates the methodology utilised in this paper; *section 5.0. Results* presents the study’s results and managerial insights inferred from these, and *section 6* provides conclusions and signposts future opportunities for further research.

2.0. The platelet supply chain: a literature review

Extant literature on platelet supply chain optimisation reviewed illustrates that international studies have critically evaluated platelet inventory management optimisation approaches from various context-specific perspectives. Although early studies extend back to the 1960s (Cohen & Pierskalla, 1979; Osorio et al., 2015), the platelet supply chain has hitherto received scant academic attention until recent years (Blake, 2017).

Haijema et al. (2009) adopted Stochastic Dynamic Programming (SDP) with simulation to arrive at policies that they claim would lead to shortage levels of less than 1% and reduce wastages from 20% to 1% during Christmas, New Year’s and Easter breaks. Due to the short shelf-life of platelets, these breaks may seriously affect the levels of shortages and wastages; making their findings particularly noteworthy. Van Dijk et al. (2009) ignored the stock age distribution and combined SDP with discrete event modelling to arrive at an ‘order-up-to’ rule for each day of the week. The major limitation of this study, however, was their

oversight in the presentation of results. Blake (2009) points out this error and strongly contests the results of Van Dijk, et al. (2009), arguing that ignoring the age-distribution of stock yields inventory levels are not practical. Further, Abdulwahab & Wahab (2014) analyzed the performance of the Canadian Blood Service and proposed a mathematical model based on approximate dynamic programming. The model examined different supply policies (circular, first-in-first-out (FIFO) and last-in-first-out (LIFO)), concluding that increasing the order delivery to two times per day and incorporating a strict FIFO policy produces minimal inventory shortages, minimum outdates and minimum average inventory. Katsaliaki & Brailsford (2017) adopted a different approach and utilised discrete-event simulation. They (*ibid*), concluded that strict adherence to FIFO inventory policy and the introduction of two routine deliveries to hospitals (vis-a-vis one), would improve the system performance, arriving at the same conclusion as Abdulwahab & Wahab (2014).

The World Health Organization (2011) guidelines recommend that adequate contingency plans should be implemented by blood centres for collection, processing and use of blood products during a pandemic. Rajendran & Srinivas (2020), consider the effects of disruption on the supply chain and adopt mathematical modelling to develop inventory management policies that offer a trade-off between wastages and shortages. Haghjoo et al. (2020); Hosseini-Motlagh et al. (2020); Hamdan & Diabat (2020) and Yaghoubi et al. (2020) made further contributions in developing solutions for optimising the supply chain under disruptions to minimize transportation, inventory and fixed costs as well as lower the rate of shortages and wastages.

3.0. Case study description

The National Health Services Blood and Transplant (NHSBT) is mandated to ensure that blood products are delivered to UK hospitals in a timely and sufficient manner. Approximately 28,000 units of blood and its components are collected weekly from fixed and

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mobile donation centres. The collected blood is then tested for ABO, rhesus grouping, infectious agents and viruses, and then processed in manufacturing centres (Baesler, et al., 2014). The three manufacturing centres within the UK (located in Manchester, Colindale and Bristol) centrifuge whole blood into the three major components: plasma, platelets and red blood cells. After processing, the blood components are stored in SHUs ready for issue when ordered by transfusion laboratory managers. The current approach for inventory management in the SHUs is to hold fixed stock targets that are reviewed every six months. This simplistic approach to inventory management of platelets tends to be time-consuming and makes it difficult to make production decisions that will result in optimal inventory levels. Since evidence-based inventory management is what leads to best practice in blood services, this study adopted one of the SHUs as a case study to illustrate the applicability of the developed simulation model and its solution approach for effective platelet inventory management.

4.0. Materials and methods

A mixed philosophical stance was adopted using interpretivism to analyse existing literature and justify the novelty of this present study (Roberts et al., 2018), and positivism (Edwards et al., 2020) to develop the simulation model. Within this overarching epistemology, a mixed-method approach was adopted, incorporating both qualitative and quantitative techniques to comprehensively model the case study. This approach was implemented in five stages that were interrelated to each other; specifically, the output of one stage iteratively informing the next as illustrated in *Figure 1*.

>> *Insert Figure 1* <<

4.1. Focus group discussions

In this multi-level single case study, a non-probability sampling technique was used to constitute a focus group. The formation of the group, discussions and site visits to two of the three manufacturing sites occurred over twenty months; an extended period that was not only

needed to negotiate research support but also secure a more comprehensive understanding of the operational activities of the SHU. This practical knowledge accrued was instrumental in developing the simulation model that accurately reflected reality. Focus group discussions were chosen because unlike interviews, they provide a more precise and clearly defined focus on the topic, enabling an interactive discussion between the participants (Pärn and Edwards, 2017). The focus group participants consisted of six senior staff members of NHSBT who are responsible for making decisions on the stock levels within the UK blood supply chain and had between 5 and 30-years' experience in blood manufacturing. Although a sample frame of ten staff was originally identified as relevant to the study, only six responded positively thus representing a 60% response rate. Semi-structured questions were used to guide interactions, and the ensuing discourse recorded, transcribed and analysed using NVivo.

4.2. Developing the simulation model

The simulation model was built on a multi-method approach combining both agent-based and discrete event modelling. This simulation model allows for the comparison of various inventory policy alternatives for the 24 different platelet products while supporting decision making in a stochastic environment by evaluating appropriate stock levels through what-if analysis (Terzi & Cavalieri, 2004). Due to the complexity of the platelet supply chain, system behaviour cannot be accurately defined (Osorio et al., 2015). However, the key variables and dependencies within the system were identified. Agent-based modelling was chosen since it allows the system's 'global' behaviour to be understood by merging individual behaviours (Grigoryev, 2018). The variable 'hospital demand' was taken as an agent interconnected with another agent; the 'SHU' and the activities within the SHU such as 'replenishment orders', 'shortages', and 'wastages' modelled using discrete-event to show how they interact within the system. The model was developed on Anylogic software together with a Java code written to replicate the daily transaction activities at the SHU.

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Overall, the simulation model considers a ‘pull’ system where the stock levels are influenced by hospital demand (refer to *Figure 2*), eliminating the need to ‘push’ blood products to the hospitals.

>> *Insert Figure 2*<<

4.2.1. Data required by the model

Data on hospital demands for the 24 different platelet categories by platelet type, quantity and blood group were obtained for the period between April 2017 and June 2019. Historical data was collected over two years to cover any seasonality that brings about variation in demand thus, presenting an accurate picture of the platelet supply chain. A source of uncertainty in this model (as replicated in a real system) is hospital demand for platelet units. Therefore, to incorporate variability in this demand, normal probability distributions at 95% confidence level were fitted to historical hospital demand data and analysed as input parameters for the model.

4.2.2. Model assumptions

The simulation was developed based on the following model assumptions: 1) Since the platelet shelf-life is seven days (Blake, 2017; Zahiri et al., 2018), the model assumes a shelf-life of four days because processing at the manufacturing centre before the product arrives at the SHU takes up to three days; 2) A FIFO ordering policy is strictly adhered to at the SHU; 3) The SHU has enough staff and equipment to operate at increased capacity; and 4) Since the model scope does not extend to the manufacturing and donation centres, it is assumed that there is enough platelet supply to replenish the SHU’s orders and the lead time is two days.

4.2.3. Model behaviour

The inventory levels of the 24 platelet types were modelled based on a continuous inventory review system; accounting for fluctuations in stock according to hospital demand and replenishment orders from the manufacturing centre (refer to *Figure 3*). Given the critical

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3 importance of having some blood products on hand (at all times) in the SHU, minimal safety
4 stock and dynamic re-order points were defined for each platelet product at a 95% service
5 level. The stock received from the manufacturing centre is stored in FIFO order at the SHU.
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12 Ideally, a hospital should be supplied with the same blood group of platelets that they have
13 ordered from the SHU but this is not always possible due to product unavailability.
14 Therefore, to manage the supply and demand, NHSBT may (in certain situations) request
15 hospitals to accept a different ABO group from what was ordered. A Negative platelets for
16 instance are the universal platelet type and are provided as a substitute for hospital demands.
17 Each hospital request is first matched with the available stock in the SHU and the model
18 allows for substitution for A Negative platelets if the same group demanded is unavailable at
19 the time of the request. To reinforce this point, Participant A said:
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30 *“In basic what I am doing all the time is making more of the more valuable units. Like*
31 *all the ones that are A Neg, I am just always making more... I am doing a lot of*
32 *substitution to push all the products out...”*
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37 The model runs for a year, which is considered to be sufficient to capture the variability in the
38 processes at the SHU - taking into consideration that one complete cycle of a platelet unit is
39 only seven days.
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45 **4.3. Model verification and validation**

46 The model was developed over four iterations based on feedback from professional
47 practitioners in the UK blood supply chain. It was first built based on the authors'
48 understanding of the platelet supply chain and thereafter modified to include ageing of the
49 platelet units over four days, A Negative platelet substitution and incorporating a dynamic re-
50 order point for each of the products. The model was verified as a true representation of the
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daily transactions at the SHU and its assumptions also viewed to be reasonable and realistic – as indicated by participant E viz:

“ The assumptions are sensible for the SHUs. ”

The model was populated with data from operations in the SHU in 2019 and was validated for reasonableness of its outputs by running the simulation 100 times. Harrell et al. (1995) highlight that this number of replications is ideal for ascertaining model output accuracy. The validation results showed no significant difference in the operational inventory and wastage levels (refer to *Table I*).

>> *Insert Table I* <<

5.0. Results

The daily inventory levels derived from the simulation were compared to the actual target stock held at the SHU for July 2019 (specifically 1st July to 28th July 2019) and the variance determined as overstocked units. Due to the short shelf-life of platelets, this study describes an overstocked unit as any surplus platelet unit held within the SHU that is nearing expiry as confirmed by participant A, viz:

“I mean if a platelet is near day 7, it is an overstock as I am very unlikely to use it, I’m pretty sure I’m gonna expire it, I have zero probability to use it”

Since the level of inventory held at the SHU influences expiries and outages, it constituted the model’s key performance indicator (KPI).

5.1. Inventory levels

Average daily inventory levels for the different platelet products were simulated based on hospital demand variations and the results determined over five iterations to capture the variability of the inputs. Simulated results were then analysed in comparison to the target inventory levels held in the SHU between 1st July 2019 and 28th July 2019. Although inventory holding costs are not considered in this model, holding higher inventory than

necessary would certainly result in higher operational costs and subsequently increased wastages and shortages. The results illustrate a 14% overstock of apheresis platelets and pooled platelets, and an 82% understock of paediatric platelets (refer to *Table 2*).

Due to the uncertainty in hospital demand for the products and their lifesaving property, the SHU holds excess inventory, especially for apheresis platelets, leading to wastages. Results indicate that by using a 'pull' system where hospital demand influences the inventory levels at the SHU, performance at the SHU is improved by reducing wastages by 78% and leading to financial savings. For instance, considering the cost implications of outdated A Negative apheresis platelets, the authors determine that the organisation loses approximately £169,915 from this one product since apheresis platelets are sold to hospitals at £232.76.

>> *Insert Table 2* <<

5.2. Outdated units

The number of outdated units is an important KPI for blood supply chains. Since blood centres face the challenge of meeting demand without overstocking, the strategy adopted involves pushing platelets out of their system to avoid internal wastages. This is implemented by moving platelets to hospitals before they expire rather than awaiting hospital orders. Although this reduces internal wastages, it does not provide a solution to the systemic problem since wastages are only transferred to another part of the supply chain and not eliminated. This common practice was explicitly stated by Participant B who said:

“When there is something that’s coming to end of shelf-life, they’ll push it into that local hospital.”

While overstocking may be a safety measure, wastage of units not only affects the operational costs of the SHU but is also a waste on the donor’s time and effort (Stanger, et al., 2012). The wastage levels at the SHU when inventory is held at the units presented in *Table 2* above were analysed in comparison to the actual monthly expiries at the SHU (refer

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to *Table 3*). Baesler, et al. (2014) maintains that an acceptable inventory policy is one that maximizes demand satisfaction while minimizing the levels of wastages. Adopting the ‘pull’ system in determining the inventory level as opposed to utilizing fixed targets as is the current case significantly reduces the overall level of expiries from 16.63% to 3% while guaranteeing the availability of stock when hospitals demand.

>> *Insert Table 3*<<

5.3. Shortages

Shortages are not experienced when executing the model because it assumes that as soon as a replenishment order is made at the SHU, the order requests will be fulfilled by the manufacturing centre within 48 hours. In reality, this may not be possible because the collection side of the supply chain affects the availability of products at the manufacturing centre. However, considering a reasonable lead time of 48 hours and implementing the ‘pull’ system the shortages are reduced from 5% to 0%.

5.4. Model resilience

The model must be resilient to unexpected challenges and disruptions in the supply chain (Sterbenz & Smith, 2010) (such as equipment breakdown and natural disasters). Thus, resilience measures in the optimisation of inventory were taken into consideration to guarantee that hospital demand for platelet products would be met in periods of unexpected changes. The simulation model’s resilience, in this case, was tested in the event of an emergency. The average hospital demands were multiplied by a factor of 10 to determine the behaviour of the model under extreme conditions. The simulation running in virtual time mode took 3 hours 24 minutes compared to the normal conditions that took 15 minutes. The results of the simulation were compared to the baseline scenario and revealed an appropriate increase in stock levels as expected and reasonable wastages. For instance, the inventory level for A Negative apheresis platelets increased to 117 units from 12 units when the

demand increased from 4 to 40 units. Wastages, on the other hand, maintained at 0 units indicating the sensitivity of the model to computing wastages relative to the product life span when there is a surge in demand.

6.0. Conclusion

Besides providing an innovative tool for inventory management at NHSBT, this study represents the first attempt to fully understand the UK platelet supply chain. A multi-method simulation model was developed to capture the activities of a real-life SHU, verified and validated by experts in the UK platelet supply chain. The performance of the SHU can be improved by keeping the daily inventory levels as (12A-, 34A+, 0AB-, 3AB+, 1B-, 8B+, 9O-, 28O+) for apheresis platelets, (7A-, 33A+, 0AB-, 1AB+, 1B-, 11B+, 7O-, 27O+) for pooled platelets and (5A-, 1A+, 0AB-, 0AB+, 0B-, 6B+, 3O-, 1O+) for paediatric platelets while strictly adhering to the FIFO policy. This paper's outcomes provide insight into the trade-off between shortages and expiry rates, and can assist decision-makers in planning for optimised inventory management. Additionally, incorporating measures of resilience into decision making guarantees the availability of platelet products under peculiar circumstances.

A major limitation of this study is that operational costs were not included in the analysis. Further, while the model represents the SHU operating independently from the activities at the manufacturing centre and collection points, in real life, the decisions made in other parts of the supply chain may significantly affect the activities at the SHU. Future research could extend this model to incorporate the activities of the other parts of the supply chain to provide a robust solution to the platelet problem experienced. In particular, incorporating variability in supply may provide a better picture of shortages as a result of variable lead times. Future research could also incorporate stock holding costs in determining the optimal stock levels to illustrate the potential savings from the implementation of the optimal stock levels.

8.0. Acknowledgements

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9.0. Declaration of interest

The authors hold no conflict of interest with the publication of the results included in this paper.

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Figure 1: Overview of the research process (Source: Authors)

Figure 2: Model pull logic (Source: Authors)

Figure 3: Model behaviour flowchart (Source: Authors)

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Table 1: Model validation- Comparison of actual data to simulation results

Apheresis platelets						
	Actual		Simulated		Variances	
Blood type	Inventory level	Wastages	Inventory level	Wastages	Inventory level	Wastages
A-	14	5	13	3	1	2
A+	35	14	34	12	1	2
AB-	1	0	1	0	0	0
AB+	3	1	3	0	0	1
B-	2	1	1	0	1	1
B+	10	5	9	5	1	0
O-	5	2	7	1	2	1
O+	26	10	27	9	1	1

Table 2: Simulation Results for Daily Inventory Levels

Blood type	Apheresis			Pooled			Paedi		
	Model results	Actual target	% over/under stock	Model results	Actual target	%over/under stock	Model results	Actual target	%over/under stock
A-	12	14	14% overstock	7	12	42% overstock	5	1	400% understock
A+	34	35	3% overstock	33	30	10% understock	1	4	75% overstock
AB-	-	1	100% overstock	-	1	100% overstock	0	1	100% overstock
AB+	3	3	0%	1	2	50% overstock	0	1	100% overstock
B-	1	2	50% overstock	1	2	50% overstock	0	1	100% overstock
B+	8	10	20% overstock	11	9	22% understock	6	1	500% understock
O-	9	5	80% understock	7	4	75% understock	3	1	200% understock
O+	28	26	8% understock	27	22	23% understock	1	3	67% overstock

Table 3: Analysis of SHU outdated (expired) units

Blood type	Apheresis			Pooled			Paedi		
	Actual units	Simulated units	% improvement	Actual units	Simulated units	% improvement	Actual units	Simulated units	% improvement
A-	5	0	100%	6	-	100%	5	0	100%
A+	14	4	71%	16	4	75%	13	0	100%
AB-	0	0	0%	0	0	0%	0	0	0%
AB+	1	0	100%	1	-	100%	2	0	100%
B-	1	0	100%	1	-	100%	1	0	100%
B+	5	1	78%	5	1	81%	6	1	83%
O-	2	1	56%	3	-	100%	3	0	100%
O+	10	4	61%	12	3	75%	14	0	100%