

Quantifying Supply Chain Resilience: A Markov Chain Approach ICTEA

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Abstract

In modern conditions, the effectiveness of supply chains depends largely on their fault tolerance. This is a complex characteristic that refers to the ability of chain participants to maintain operational efficiency or restore the performance of technological processes under the negative impact of external factors. It is crucial for supply chain participants to be adequately prepared to respond to these factors to avoid significant losses for businesses, markets, and national economies as a whole. The aim of our research is to establish a methodology for qualitatively assessing the resilience of supply chains. This methodology will allow us to obtain precise resilience values, supporting decision-making in response to deviations and failures within supply chains. The research examines contemporary methods for identifying the causes of supply chain failures, studies the dynamics of process states in individual links of supply chains, develops a methodology for assessing their fault tolerance based on the theory of Markov processes, and provides recommendations for improving the fault tolerance of supply chains. The methodology was used to evaluate the fault tolerance of consumer goods supplies from manufacturers to the retail network. The supply chain parameters were adjusted based on the obtained values of fault tolerance, resulting in increased sales for the trading company.

Keywords: Supply Chain Resilience, Fault Tolerance, Markov Chain Approach, Supply Chain Failures, Risk Management, Operational Efficiency, Decision-Making Support, Process State Dynamics, Resilience Assessment, External Factors Impact, Retail Supply Chains, Supply Chain Optimization, Business Continuity, Performance Recovery, Stochastic Processes.

1. Introduction

In the most general sense, the term "Supply Chains Resilience" (SCR) characterises the ability of supply chain participants to maintain operational efficiency or quickly restore normal operation in case of process failures [1,2,3,4]. Resilient execution of operations by supply chain participants is one of the most important conditions for the effective functioning of modern Global Value Chains (hereinafter - "GVC") and sustainable development of the economies of individual countries. The participation of national companies in GVCs positively affects social well-being and business efficiency by creating new jobs, attracting investment in infrastructure, and increasing foreign trade volumes [5,6,7,8].

Recently, SC failures have become more frequent, resulting in increased financial losses for GVC participants around the world, rising inflation rates, and deterioration of public welfare in many countries [9,10, 11]. For example, losses to world trade from the blockage of the Suez Canal by the grounded container ship "Ever Given" reached \$10 billion per week. During the COVID-19 pandemic, many assembly plants around the world stopped operations due to the disruption of component supplies [12-20]. These

examples highlight the practical significance of research in supply chain fault tolerance (Resilient functioning of SC). The results of these studies allow to increase the efficiency of response to failures and reduce the severity of their unfavourable consequences for SC participants [9,21,22,23]. The key research objective is to develop a methodology for the qualitative assessment of SCR (Quantification of Supply Chain Resilience, hereinafter - QSCR). The research will consider modern approaches to identifying the causes of SC failures, the influence of random factors on the change of SC states, will propose a method of quantitative assessment of fault tolerance, as well as the basic rules for making managerial decisions that ensure fast and efficient restoration of SC links operation.

2. Materials and methods

The SCR problem is an important part of modern scientific research in logistics [24, 25, 7, 26,27,28,29,30]. According to the authors [31], the basic definition of the term SCR was formulated in [32] - it is the ability of SC to return to the initial state or move to a new, more desirable state, after a disruption in one or more links of the chain. In the context of this study, this definition should be supplemented with three characteristics.

In the study [33,34] it is noted that SC operation is associated with many risks, as they are open complex-structured dynamic systems that actively interact with the external environment. This interaction is manifested in the following features of the impact of external factors on SCRs.

Let us introduce notations of SC links and their states. Under the state we will understand the result of the technological process execution ("successfully" or "unsuccessfully" executed).

The value of "transient" probabilities P_{ij} depends on two parameters: firstly, on the "strength" $q_{ij}(z)$ of the impact of some unfavourable external factor on the operation of SC links, and secondly, on the "frequency" $v_{ij}(z)$ of the manifestation of this factor [35].

"Frequency of occurrence of unfavourable impacts" $v_{ij}(z)$ is determined by the nature of external factors. For example, in some regions natural disasters occur more often than in others. As a rule, statistical data are used to assess the value of the "frequency" of manifestation of external factors on individual processes, and expert opinions are additionally taken into account when assessing the "strength" of the influence of external factors. The parameters $q_{ij}(z)$ and $v_{ij}(z)$ can take values from 0 to 1. If at least one parameter is zero, the undesired event E_{ij} does not occur, respectively the system does not go from S_i to S_j .

An undesirable event E_{ij} can occur with probability P_{ij} , only when both parameters are greater than zero: $q_{ij}(z) > 0$; $v_{ij}(z) > 0$. Therefore, the final value of the probability P_{ij} of the system transition from the state S_i to the state S_j under the influence of unfavourable external factors is defined as the product of:

$$P_{i,j} = q_{i,j}(z) \cdot v_{i,j}(z). \quad (1)$$

The sum of transition probabilities P_{ij} for each i -th event must be equal to one, since, whatever state the system is in before the next transition, all subsequent j -th events are incompatible and form a complete group:

$$\sum_{j=1}^n P_{i,j} = 1. \quad (2)$$

The probability values $p_i^*(k)$ and $p_i(k)$ are calculated using the same methodology based on the provisions of the theory of Markov processes.

3. RESULTS AND DISCUSSION

Markov Process is used as a means of creating a dynamic model of stochastic processes that are realised in different systems [36-42]. In this study, the SC model is designed as a homogeneous Markov Chain (homogeneous Markov Chain) with discrete time, in which transitions of the system from state to state are possible only at strictly defined, pre-fixed moments of time: t_1, t_2, \dots . In the time intervals between these moments, the system S preserves its state [43-48].

Based on the modelling results, QSCR will be produced in the form of probability values $p_i^*(k)$ of finding the system in successful intermediate (S_2^*, S_5^*, S_{10}^*) and final (S_{12}^*) states. In turn, the obtained values of probabilities $p_i(k)$ of being in "unsuccessful" states will be used in the selection of corrective solutions that ensure the increase of $p_i^*(k)$.

After corrective solutions, the transition probabilities P_{ij} may be revised. However, this fact does not indicate that the SC model will become an inhomogeneous Markov chain. After corrective solutions, a new SC model is created, which will also be a "homogeneous Markov chain".

SC functioning in random conditions is characterised by the change of link states after each step. Transient probabilities of SC states after the k -th step can be written as conditional probabilities of preceding and following events:

$$P_{ij} = P(S_j^{(k)} / S_i^{(k-1)}). \quad (4)$$

To calculate $p_i(k)$, we write the transition probabilities P_{ij} as a rectangular matrix:

$$\|P_{ij}\| = \begin{vmatrix} 0 & 0.7 & 0.1 & 0.2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.7 & 0.2 & 0.1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0.95 & 0.05 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.8 & 0.2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1.0 & 0 & 0 & 0 & 0 \\ 0.3 & 0 & 0.6 & 0.1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.8 & 0.2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1.0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1.0 \\ 1.0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{vmatrix} \quad (3)$$

$$\begin{bmatrix} 1.0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

The state probabilities $p_i(k)$ after each k-th step will be determined by the dynamics of the possible states and the values of the transition probabilities for the corresponding pairwise states. According to the methodology from [43], the probabilities of states $p_i(k)$ are determined by the recurrence formula:

$$p_i(k) = \sum_{j=1}^n (p_j(k-1) \cdot P_{ji}). \quad (5)$$

We will record the obtained values of state probabilities $p_i(k)$ for the next step in Table 1.

Table 1. Probabilities $p_i(k)$ of SC link states at the k-th step (developed by the authors only).

S_i	1	2	3	4	5	6	7	8	9	10	11	12	$p(k)$	$R_i'(k)$
Step 0	1	0	0	0	0	0	0	0	0	0	0	0	1	1
Step 1	0	0.7*	0.1	0.2	0	0	0	0	0	0	0	0	1	0.7
Step 2	0.1	0.19	0.01	0	0.49*	0.14	0.07	0	0	0	0	0	1	0.49
Step 3	0.031	0.07	0.052	0.027	0.133	0.038	0.019	0.14	0	0.392*	0.098	0	1	0.392
Step 4	0.1557	0.04735	0.01585	0.0081	0.049	0.014	0.007	0.038	0.112	0.1344	0.0266	0.392*	1	0.392

* fault tolerance value for a "successful" state

Step 0 ($k = 0$, SC start). At the initial moment ($k = 0$) the system is in the state S_1 . The probability of the system being in this state is $p_1(0) = 1$. The probabilities of all other states are zero.

Step 1 ($k = 1$). After the first step, the system can move to states S_2 , S_3 and S_4 with transition probabilities P_{12} , P_{13} and P_{14} , which are written in the first row of the matrix. Then the probabilities of the states that SC can be in after the first step are as follows: probability of SC staying in state $S_2^{(1)}$: $p_2(1) = 1 \cdot 0.7 = 0.7$ (**successful process**); probability of SC staying in state $S_3^{(1)}$: $p_3(1) = 1 \cdot 0.1 = 0.1$ (unwanted deviation); probability of SC staying in state $S_4^{(1)}$: $p_4(1) = 1 \cdot 0.2 = 0.2$ (unwanted deviation).

The sum of the probabilities of SC staying in states $S_2^{(1)}$, $S_3^{(1)}$ and $S_4^{(1)}$ after the first step is equal to one: $p(1) = 1.0$:

$$p(k) = \sum_{i=1}^n p_i(k). \quad (6)$$

Step 2 ($k = 2$). After the second step, the SC can move from each possible state S_2 , S_3 and S_4 to the following states: from state S_2 to states S_5 , S_6 and S_7 ; from state S_3 to state S_1 ; from state S_4 to states S_2 and S_3 . Let us determine the probabilities of SC staying in the listed states - S_1 , S_2 , S_3 , S_{235} , S_6 and S_7 after the second step: probability of SC staying in state $S_1^{(2)}$: $p_1(2) = p_3(1) \cdot P_{31} = 0.1 \cdot 1 = 0.1$ (**resumption of supply after failures**); probability of SC staying in state $S_2^{(2)}$: $p_2(2) = p_4(1) \cdot P_{42} = 0.2 \cdot 0.95 = 0.19$ (**retrying a successful process**); probability of SC staying in state $S_3^{(2)}$: $p_3(2) = p_4(1) \cdot P_{43} = 0.2 \cdot 0.05 = 0.01$ (unwanted bias); probability of SC staying in state $S_5^{(2)}$: $p_5(2) = p_2(1) \cdot P_{25} = 0.7 \cdot 0.7 = 0.49$

(**successful process**); probability of SC staying in state $S_6^{(2)}$: $p_6(2) = p_2(1) \cdot P_{26} = 0.7 \cdot 0.2 = 0.14$ (unwanted bias); probability of SC staying in state $S_7^{(2)}$: $p_7(2) = p_2(1) \cdot P_{27} = 0.7 \cdot 0.1 = 0.07$ (unwanted bias). The sum of the probabilities of SC staying in states S_1 , S_2 , S_3 , S_{235} , S_6 and S_7 after the second step is equal to one: $p(2) = 1.0$.

Step 3 ($k = 3$). After the third step, the SC can move from each previous state S_1 , S_2 , S_3 , S_{235} , S_6 and S_7 to the following states: from state S_1 to states S_2 , S_3 and S_4 ; from state S_2 to states S_5 , S_6 and S_7 ; from state S_3 to state S_1 ; from state S_5 to states S_{10} and S_{11} ; from state S_6 to state S_8 ; from state S_7 to states S_1 , S_3 and S_4 .

The sum of the probabilities of SC staying after the third step in states S_1 , S_2 , S_3 , S_{34} , S_5 , S_{56} , S_7 , S_{78} , S_{10} and S_{11} is equal to one: $p(3) = 1.0$.

Step 4 ($k = 4$). After the third step, the SC can move from each previous state S_1 , S_2 , S_3 , S_{235} , S_4 , S_5 , S_{56} , S_7 , S_{78} , S_{10} and S_{11} to all possible states: from state S_1 to states S_2 , S_3 and S_4 ; from state S_2 to states S_5 , S_6 and S_7 ; from state S_3 to state S_1 ; from state S_4 to states S_2 and S_3 ; from state S_5 to states S_{10} and S_{11} ; from state S_6 to state S_8 ; from state S_7 to states S_1 , S_3 and S_4 ; from state S_8 to states S_9 and S_{10} ; from state S_{10} to state S_{12} ; from state S_{11} to state S_1 .

Let us determine the probabilities of SC staying in states S_1 , S_2 , S_3 , S_4 , S_5 , S_6 , S_7 , S_{789} , S_{10} , S_{11} and S_{12} after the fourth step: probability of SC being in state $S_1^{(4)}$: $p_1(4) = p_3(3) \cdot P_{31} + p_7(3) \cdot P_{71} + p_{11}(3) \cdot P_{11,1} = 0.052 \cdot 1 + 0.019 \cdot 0.3 + 0.098 \cdot 1.0 = 0.1557$ (**resumption of supply after failures**); probability of SC being in state $S_2^{(4)}$: $p_2(4) = p_1(3) \cdot P_{12} + p_4(3) \cdot P_{42} = 0.031 \cdot 0.7 + 0.027 \cdot 0.95 = 0.04735$ (**retry success of the process**); probability of SC being in state $S_3^{(4)}$: $p_3(4) = p_1(3) \cdot P_{13} + p_4(3) \cdot P_{43} + p_7(3) \cdot P_{73} = 0.031 \cdot 0.1 + 0.027 \cdot 0.05 + 0.019 \cdot 0.6 = 0.01585$ (undesirable variance); probability of SC being in state $S_4^{(4)}$: $p_4(4) = p_1(3) \cdot P_{14} + p_7(3) \cdot P_{74}$

$= 0.031 \cdot 0.2 + 0.019 \cdot 0.1 = 0.0081$ (unwanted variance);
 probability of SC staying in state $S_5^{(4)}$: $p_5(4) = p_2(3) \cdot P_{2,5}$
 $= 0.07 \cdot 0.7 = 0.049$ (**retrying a successful process**);
 probability of SC staying in state $S_6^{(4)}$: $p_6(4) = p_2(3) \cdot P_{2,6}$
 $= 0.07 \cdot 0.2 = 0.014$ (unwanted bias); probability of SC
 staying in state $S_7^{(4)}$: $p_7(4) = p_2(3) \cdot P_{2,7} = 0.07 \cdot 0.1 =$
 0.007 (unwanted bias); probability of SC staying in state
 $S_8^{(4)}$: $p_8(4) = p_6(3) \cdot P_{6,8} = 0.038 \cdot 1.0 = 0.038$ (unwanted
 bias); probability of SC staying in state $S_9^{(4)}$: $p_9(4) = p_8$
 $(3) \cdot P_{8,9} = 0.14 \cdot 0.8 = 0.112$ (unwanted bias); probability
 of SC being in state $S_{10}^{(4)}$: $p_{10}(4) = p_5(3) \cdot P_{5,10} + p_8(3) \cdot$
 $P_{8,10} = 0.133 \cdot 0.8 + 0.14 \cdot 0.2 = 0.1344$ (**retry success of**
the process); probability of SC staying in state $S_{11}^{(4)}$: p_{11}
 $(4) = p_5(3) \cdot P_{5,11} = 0.133 \cdot 0.2 = 0.0266$ (unwanted bias);
 probability of SC staying in state $S_{12}^{(4)}$: $p_{12}(4) = p_{10}(3) \cdot$
 $P_{10,12} = 0.392 \cdot 1.0 = 0.392$ (**successful completion of**
the process). The sum of the probabilities of SC staying
 in all twelve states after the fourth step is equal to one: $p(4)$
 $= 1.0$. If we continue modelling the SC processes, in the
 fifth step new transitions between states will be added that
 will influence the final fault tolerance score in state S_{12} .

The value of the proposed QSCR methodology lies in
 the fact that it allows us to obtain generalised quantitative
 assessments of reliability of operational activities of all SC
 participants (links): fault tolerance indices $R_i(k)$ and R and
 probability of staying in unsuccessful states $p_i(k)$. The
 calculated values of $R_i(k)$ for "successful" states show that
 with each "step" this indicator decreases and eventually
 reaches the value $R = 0.392$. This information suggests
 that at the current level of failure preparedness of SC
 participants, only 39% of orders will be fulfilled as planned.
 The rest will be delayed or cancelled. It is possible to
 increase the SCR by reducing the transition probabilities
 $P_{i,j}$ to unsuccessful states $p_i(k)$ through response
 measures [37, 33, 48, 49, 50, 51, 52, 53, 54, 55, 56, 57].

References

1. Finance and Risk Management for International Logistics and the Supply Chain. Editor(s): Stephen Gong, Kevin Cullinane. Elsevier. 2018. – <https://doi.org/10.1016/C2017-0-00048-4>.
2. Yosef Sheffi. The Resilient Enterprise: Overcoming Vulnerability for Competitive Advantage. MIT Press, 2005. 338 p.
3. 2021 Global Risk Management Survey // Aon plc., 2021. – 142 p. https://assets.foleon.com/euwest2/uploads/7e3kk3/48136/aon2021_global_risk_management_survey_findings.8ca75da55c29.pdf.
4. Supply Chain Disruption // Gartner, Inc. – 2023. – June 22 – <https://www.gartner.com/en/supply-chain/insights/supply-chain-disruption>.
5. Kano, L., Tsang, E.W.K. & Yeung, H.Wc. Global value chains: A review of the multi-disciplinary literature. J Int Bus Stud 51, 577–622 (2020). <https://doi.org/10.1057/s41267-020-00304-2>.
6. Konina, N. (2023). Global Value Chains. In: Bulatov, A. (eds) World Economy and International Business. Contributions to Economics. Springer, Cham. https://doi.org/10.1007/978-3-031-20328-2_30.

3. Conclusion

The methodology's implementation begins with
 creating a model of changes in the states of technological
 processes of SC participants [58]. To achieve this, we
 collect detailed information on failures that affect their
 technological processes, including how often and under
 what negative scenarios 'unsuccessful' events develop.
 We then analyse all unsuccessful states and identify the
 factors that determine the outcome of processes in all SC
 links. The parameters for replenishing products delivered
 with failure conditions are adjusted based on the obtained
 results. Additionally, the size of insurance stocks for these
 products in distribution centres and retail networks is
 modified. These measures increase fault tolerance and
 reduce losses from late deliveries. The effect is achieved
 by reducing the range of goods that are periodically
 unavailable in the retail network due to late deliveries.
 Additionally, the average duration of time during which the
 goods are unavailable is reduced, resulting in increased
 sales of such goods.

This example demonstrates the methodological
 commonality and connection between QSCR
 methodology and procurement planning methods.
 Therefore, it appears promising to conduct research on
 implementing SCR indicators into existing methods for
 planning and optimizing business processes in the supply
 chain. Additionally, given the large amount of data
 generated in modern supply chains, applying machine
 learning methods for quantitative assessment of SCR
 indicators is a promising area for further research.

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7. Katsaliaki, K., Galetsi, P. & Kumar, S. Supply chain disruptions and resilience: a major review and future research agenda. Ann Oper Res 319, 965–1002 (2022). <https://doi.org/10.1007/s10479-020-03912-1>.
8. Aubin J. P., Bayen A. M., Saint-Pierre P. Viability theory: new directions. – Springer Science & Business Media, 2011.
9. Parast, M.M., Shekarian, M. (2019). The Impact of Supply Chain Disruptions on Organizational Performance: A Literature Review. In: Zsidisin, G., Henke, M. (eds) Revisiting Supply Chain Risk. Springer Series in Supply Chain Management, vol 7. Springer, Cham. https://doi.org/10.1007/978-3-030-03813-7_21.
10. Revilla, E., Acero, B., Sáenz, M.J. (2023). Resilience in the Supply Chain. In: Sarkis, J. (eds) The Palgrave Handbook of Supply Chain Management. Palgrave Macmillan, Cham. https://doi.org/10.1007/978-3-030-89822-9_106-1.
11. Alimova, Z., Niyazova, G., & Sabirova, D. (2024). Specifications for applying high octane additives to the motor gasoline in hot climate. In E3S Web of Conferences (Vol. 508, p. 07010). EDP Sciences.
12. Handfield, R.B., Graham, G. and Burns, L. (2020), "Corona virus, tariffs, trade wars and supply chain evolutionary design", International Journal of Operations &

Production Management, Vol. 40 No. 10, pp. 1649-1660. <https://doi.org/10.1108/IJOPM-03-2020-0171>.

13. Remko, v.H. (2020), "Research opportunities for a more resilient post-COVID-19 supply chain – closing the gap between research findings and industry practice", *International Journal of Operations & Production Management*, Vol. 40 No. 4, pp. 341-355. <https://doi.org/10.1108/IJOPM-03-2020-0165>.

14. Larin O.N., Novikov A.N. (2023) Problemy organizatsii dostavki slaboalkogolnoy produktsii v sovremennykh usloviyakh [Problems of delivery of low-alcohol products in modern conditions]. *Ekonomika, predprinimatelstvo i pravo*. 13. (4). – 1199-1210. doi: 10.18334/epp.13.4.117478.

15. Kononova K.M. (2023) Kak sanktsii vliyayut na ekonomiku strany? [How do sanctions affect the country's economy?]. *Ekonomika i sotsium: sovremennye modeli razvitiya*. 13. (1). – 25–32. doi: 10.18334/ecsoc.13.1.117701.

16. Khakimov, R. Ayrapetov, D. Anti-corrosion coatings for the protection of metal parts of transport equipment AIP Conference Proceedings, 2024, 3045(1), 030065 <https://www.scopus.com/record/display.uri?eid=2-s2.0-85188443782&origin=resultslist>

17. Amiti, Mary, Stephen J. Redding, and David E. Weinstein. 2019. "The Impact of the 2018 Tariffs on Prices and Welfare." *Journal of Economic Perspectives*, 33 (4): 187-210. DOI: 10.1257/jep.33.4.187.

18. Pablo D Fajgelbaum, Pinelopi K Goldberg, Patrick J Kennedy, Amit K Khandelwal, The Return to Protectionism, *The Quarterly Journal of Economics*, Volume 135, Issue 1, February 2020, Pages 1–55, <https://doi.org/10.1093/qje/qjz036>.

19. Tu, X., Du, Y., Lu, Y. et al. US-China Trade War: Is Winter Coming for Global Trade?. *J OF CHIN POLIT SCI* 25, 199–240 (2020). <https://doi.org/10.1007/s11366-020-09659-7>.

20. Anton Pichler et al. Building an alliance to map global supply networks. *Science* 382, 270-272(2023). DOI:10.1126/science.adi7521.

21. Liu, H., Han, Y. & Zhu, A. Modeling supply chain viability and adaptation against underload cascading failure during the COVID-19 pandemic. *Nonlinear Dyn* 110, 2931–2947 (2022). <https://doi.org/10.1007/s11071-022-07741-8>.

22. Shuhan Meng, Xianhua Wu. Risk analysis of cruise ship supply chain based on the set pair analysis-Markov chain model // *Ocean & Coastal Management*. – 2023. – Vol. 245. – 106855. – <https://doi.org/10.1016/j.ocecoaman.2023.106855>.

23. Vassiliou, P.-C.G. 2021. "Non-Homogeneous Markov Set Systems" *Mathematics* 9, no. 5: 471. <https://doi.org/10.3390/math9050471>.

24. Anbumozhi, V., Kimura, F., Thangavelu, S.M. (2020). Global Supply Chain Resilience: Vulnerability and Shifting Risk Management Strategies. In: Anbumozhi, V., Kimura, F., Thangavelu, S. (eds) *Supply Chain Resilience*. Springer, Singapore. https://doi.org/10.1007/978-981-15-2870-5_1.

25. Garcia Herrero, A. 2023. Resilience of Global Supply Chain: Facts and Implications. ADBI Working Paper 1398. Tokyo: Asian Development Bank Institute. Available: <https://doi.org/10.56506/UKPK2510>.

26. Golan, M.S., Jernegan, L.H. & Linkov, I. Trends and applications of resilience analytics in supply chain modeling: systematic literature review in the context of the

COVID-19 pandemic. *Environ Syst Decis* 40, 222–243 (2020). <https://doi.org/10.1007/s10669-020-09777-w>.

27. Matosov M.V., Larin O.N. (2023) Organizatsiya vzaimodeystviya uchastnikov farmatsevticheskoy tsepochki postavok [Interaction of participants in the pharmaceutical supply chain]. *Ekonomika, predprinimatelstvo i pravo*. 13. (4). – 983–994. doi: 10.18334/epp.13.4.117492.

28. Jafar Razmi, Alireza Taheri Moghadam, Fariborz Jolai. An Evaluative Continuous Time Markov Chain Model for a Three Echelon Supply Chain with Stochastic Demand and Lead Time // *IFAC-PapersOnLine*. 2015. – Vol. 48. – Is. 3. – pp. 248-253. – <https://doi.org/10.1016/j.ifacol.2015.06.089>.

29. What is supply chain? // McKinsey. – August 2022. – 6 p. – URL: https://www.mckinsey.com/~media/mckinsey/featured%20insights/mckinsey%20explainers/what%20is%20supply%20chain/what_is_supply_chain.pdf.

30. Umirov, N., Abdurokhmonov, S., Ganiboyeva, E., & Alimova, Z. (2024). Thermal equilibrium of the tractor and vehicle engines' cooling systems in agriculture technological processes. In *BIO Web of Conferences* (Vol. 105, p. 05020). EDP Sciences.

31. Chowdhury, M.M.H. and Quaddus, M. (2016), "Supply chain readiness, response and recovery for resilience", *Supply Chain Management*, Vol. 21 No. 6, pp. 709-731. <https://doi.org/10.1108/SCM-12-2015-0463>.

32. Christopher, M. and Peck, H. (2004), "Building the Resilient Supply Chain", *The International Journal of Logistics Management*, Vol. 15 No. 2, pp. 1-14. – <https://doi.org/10.1108/09574090410700275>.

33. Qun Wu, Jiayi Zhu, Yang Cheng. The effect of cross-organizational governance on supply chain resilience: A mediating and moderating model // *Journal of Purchasing and Supply Management*. – 2023. – Vol. 29. – Is. 1. – 100817. – <https://doi.org/10.1016/j.pursup.2023.100817>.

34. Mohammad Zarei, Mohammad Bagheri, Payman Dehghanian. Markov-chain-driven optimization of inspection-based maintenance, Part I: Models and methods // *Electric Power Systems Research*. – 2024. – Vol. 228. – 110049. – <https://doi.org/10.1016/j.epsr.2023.110049>.

35. Fozilov, G., Islam, R., Akhmedov, A., Nulloev, U., Shodmonov, S., Alimova, Z., & Yuldoshev, S. (2024). Results of theoretical and experimental researches about determination the corn seed separator sieve parameters of the corn-thresher machine. In *BIO Web of Conferences* (Vol. 105, p. 04009). EDP Sciences.

36. Mark A. Pinsky, Samuel Karlin. An Introduction to Stochastic Modeling (Fourth Edition), Editor(s): Mark A. Pinsky, Samuel Karlin. Academic Press, 2011, pp. 79-163. – <https://doi.org/10.1016/B978-0-12-381416-6.00003-4>.

37. Hernández-Espallardo, M., Rodríguez-Orejuela, A. and Sánchez-Pérez, M. (2010), "Inter-organizational governance, learning and performance in supply chains", *Supply Chain Management*, Vol. 15 No. 2, pp. 101-114. <https://doi.org/10.1108/13598541011028714>.

38. Martínez-Zarzoso, I. 2023. Trade Facilitation and Global Value Chains in a Post-Pandemic World. ADBI Working Paper 1378. Tokyo: Asian Development Bank Institute. Available: <https://doi.org/10.56506/YUCF1465>.

39. Ponomarev S. Y., Holcomb M. C. Understanding the concept of supply chain resilience // *The international journal of logistics management*. – 2009. – T. 20. – №. 1. – C. 124-143.

40. Procter & Gamble warns that coronavirus will have material impact on March quarter earnings – <https://www.marketwatch.com/story/procter-gamble-warns-that-coronavirus-will-have-material-impact-on-march-quarter-earnings-2020-02-20>.
41. Brandon-Jones, E, Squire, B, Autry, CW & Petersen, KJ 2014, 'A Contingent Resource-Based Perspective of Supply Chain Resilience and Robustness', *Journal of Supply Chain Management*, vol. 50, no. 3, pp. 55-73. <https://doi.org/10.1111/jscm.12050>.
42. Jüttner, U. and Maklan, S. (2011), "Supply chain resilience in the global financial crisis: an empirical study", *Supply Chain Management*, Vol. 16 No. 4, pp. 246-259. <https://doi.org/10.1108/13598541111139062>.
43. Venttsel E.S. Issledovanie operatsiy [Operations Research]. Moscow, Sovetskoe radio Publ., 1972, 552 p. (In Russian)
44. Hakimov, R. Ayratov, D. Regeneration of spent low-freezing liquids AIP Conference Proceedings, 2024, 3045(1), 030064 <https://www.scopus.com/record/display.uri?eid=2-s2.0-85188463131&origin=resultslist>
45. Sá, M.M.d., Miguel, P.L.d.S., Brito, R.P.d. and Pereira, S.C.F. (2020), "Supply chain resilience: the whole is not the sum of the parts", *International Journal of Operations & Production Management*, Vol. 40 No. 1, pp. 92-115. <https://doi.org/10.1108/IJOPM-09-2017-0510>.
46. Duan Liu, Qiuhong Wang, Aidi Wang, Shujie Yao. Export profitability and firm R&D: on China's export diversification under trade war // *Structural Change and Economic Dynamics*. – 2023. – Vol. 67. – pp. 151-166. – <https://doi.org/10.1016/j.strueco.2023.07.012>.
47. Oleg Larin, Dmitry Tarasov, Leonid Mirotn, Vladimir Rublev, Denis Kapski. Resilient Supply Chain Management Model. SHS Web Conf., 93 (2021) 03005. DOI: <https://doi.org/10.1051/shsconf/20219303005>.
48. Yi Liu, Yadong Luo, Ting Liu. Governing buyer-supplier relationships through transactional and relational mechanisms: Evidence from China // *Journal of Operations Management*. – 2009. – Vol. 27. – Is. 4. – pp. 294-309. – <https://doi.org/10.1016/j.jom.2008.09.004>
49. Shadimetov, Y. Ayratov, D. Transport ecology in the context of the coronavirus pandemic E3S Web of Conferences, 2024, 497, 02042
50. Vasco M Carvalho, Makoto Nirei, Yukiko U Saito, Alireza Tahbaz-Salehi, Supply Chain Disruptions: Evidence from the Great East Japan Earthquake, *The Quarterly Journal of Economics*, Volume 136, Issue 2, May 2021, Pages 1255–1321, <https://doi.org/10.1093/qje/qjaa044>.
51. For Resilient Supply Chains, Think Local – <https://www.sap.com/cis/insights/viewpoints/for-resilient-supply-chains-think-local.html>.
52. Risk, resilience, and rebalancing in global value chains / Lund S. and others // McKinsey Global Institute. – August 2020. – 99 p. – <https://www.mckinsey.com/~media/mckinsey/business%20functions/operations/our%20insights/risk%20resilience%20and%20rebalancing%20in%20global%20value%20chains/risk-resilience-and-rebalancing-in-global-value-chains-full-report-vh.pdf>.
53. The effect comparative evaluation of energy-saving additives on the bitumen properties, Alimjon Riskulov, Takhira Sidikova, Rovshan Khakimov and Jamshed Avliyokulov, E3S Web of Conf., 401 (2023) 05028 DOI: <https://doi.org/10.1051/e3sconf/202340105028>
54. Xuanning Ji, Minghai Chen, Zhengheng Pu, Yanfu Fu, Tao Tao, Kunlun Xin. Markov decision process based value chain calculation of water distribution network scheduling // *Water-Energy Nexus*. – 2024. – Vol. 7. – pp. 13-25. – <https://doi.org/10.1016/j.wen.2023.12.001>.
55. Reprioritizing Risk and Resilience for a Post-COVID-19 Future. Detailed insights and analysis from Aon's COVID-19 Risk Management and Insurance Survey // Aon plc., 2021. – 50 p. – https://assets.foleon.com/eu-west-2/uploads-7e3kk3/48136/covid-19_risk_management_insurance_survey_v7.55f9f69bd6f0.pdf.
56. Barkhanadzhyan, A.L., Khakimov, R.M., Ibragimov, B.D., Sobirova, D.K., Abdukarimova, G.U., Ayratov, D.A. Problem of using waste of paint and varnishing materials and their disposal (2020) *Bulletin of the Tomsk Polytechnic University, Geo Assets Engineering*, 331 (9), pp. 179-185.
57. Mitigate Supply Chain Risk with Trade Disruption Insurance – November 2022 – <https://www.aon.com/insights/articles/2022/mitigate-supply-chain-risk-with-trade-disruption-insurance>.
58. Razzokov, T. X., Toshtemirov, S. J., Ergashov, G. K., Kiyamov, A. Z., Rashidov, N. S., & Alimova, Z. X. (2024). Substantiation of the results of uneven leveling of heaps on the dryer conveyor. In *E3S Web of Conferences* (Vol. 494, p. 04048). EDP Sciences.