



IJRASET

International Journal For Research in
Applied Science and Engineering Technology



INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 13 **Issue:** V **Month of publication:** May 2025

DOI: <https://doi.org/10.22214/ijraset.2025.71125>

www.ijraset.com

Call:  08813907089

E-mail ID: ijraset@gmail.com

An Economic Production Quantity Model Approach by Lowering the Impact of Carbon Emission under Uncertain Environment

V Vinola Elisabeth¹, S Rexlin Jeyakumari²

¹Research Scholar, PG and Research Department of Mathematics, Holy Cross College (Autonomous), Affiliated to Bharathidasan University, Tiruchirappalli, 620002, Tamil Nadu, India

²Assistant Professor of Mathematics, PG and Research Department of Mathematics, Holy Cross College (Autonomous), Affiliated to Bharathidasan University, Tiruchirappalli, 620002, Tamil Nadu, India

Abstract: The primary factor of rising global temperatures is the increase in carbon dioxide emissions, which have serious consequences on both ecological and social well-being worldwide. It contributes to environmental damage and poses various health hazards to society. This study aims to develop a fuzzy economic production quantity model with emission reduction strategies to promote economic profitability and ecological sustainability in inventory functions. Two models were developed: the fuzzy model and the crisp model. The parameters in this research work are taken as trapezoidal fuzzy numbers. The Yager ranking method and beta distribution methods are used for the solution procedure. A numerical example is used for a better understanding of the research work.

Keywords: Carbon emission, Sustainability, EPQ, Fuzzy, Green technology investment

I. INTRODUCTION

As global warming becomes a serious issue, several governments are increasingly pursuing carbon emission reduction throughout all phases of the inventory process. For many businesses, controlling inventory decisions while reducing waste is a crucial task. To address this, governments adopt various strategies, such as imposing carbon taxes at different production stages, setting emission limits, promoting recycling, and encouraging green investments. Recycling is widely regarded as an indispensable and globally acknowledged strategy for mitigating atmospheric pollution. It plays a vital role in protecting us from impending ecological catastrophes. Recycling consumes considerably less energy and also assists in protecting the planet's precious natural assets than producing stuff from scratch, which takes up plenty of energy. Some of the recent research investigations include Dwicahyani et.al (2017) proposed an inventory model with limitations in remanufacturing cycles and broader sustainability issues. In (2018), Daryanto et.al in their work presented a fundamental sustainable EPQ model that takes account of carbon emissions from waste disposal, warehousing, and production activities. Priyan et.al (2022) developed a manufacturing inventory model that incorporates emissions from transportation, production, and storage activities, which are meant to be decreased by green investment. In (2023), Shah et.al presented a production inventory model with manufacturing and inventory factors for deterioration with green technology investment. Singh et.al (2024) formulated an inventory model by incorporating machine learning for handling imperfect goods under green investment technology. Price shifts in many items on the marketplace result in an immense degree of ambiguity. The primary purpose of fuzzy set theory is the mathematical discussion of imprecision and unpredictability. Maity et.al (2008) offered a fuzzy environment optimum control recycling production inventory structure. Shekarian et.al (2016) established a reverse inventory model that includes imprecise demand and restoration rates for products that can be recovered, as well as the implications of the learning hypothesis on the recoverable manufacturing cycle. In this research study, Manufacturing inventory model with recycling and green investment is considered. Unpredictable variables in fuzzy approach are expressed by trapezoidal fuzzy numbers. Two methods were employed in fuzzy approach: Yager ranking and the beta distribution method.

II. DEFINITIONS

1) Definition 1 (Fuzzy set):

A fuzzy set \tilde{B} in a universe of discourse X is defined as the following set of pairs $\tilde{B} = \{(x, \mu_{\tilde{B}}(x)) : x \in X\}$. Here $\mu_{\tilde{B}} : X \rightarrow [0,1]$ is a mapping called the membership value of $x \in X$ in a fuzzy set \tilde{B} .

2) Definition 2 (Trapezoidal fuzzy number):

A trapezoidal fuzzy number $\tilde{B}_M = (m_1, m_2, m_3, m_4)$ is represented with membership function $\mu_{\tilde{B}_M}$ as

$$\mu_{\tilde{B}_M}(j) = \begin{cases} 0, & j \leq m_1; \\ \frac{j - m_1}{m_2 - m_1}, & m_1 < x \leq m_2; \\ 1, & x = m_2 \\ \frac{m_3 - j}{m_4 - m_3}, & m_3 < j \leq m_4; \\ 0, & \text{otherwise} \end{cases}$$

3) Definition 3 (Arithmetic operations): The arithmetic operations between Trapezoidal fuzzy numbers proposed are given below.

Let us consider, $\tilde{U} = [u_1, u_2, u_3, u_4]$ and $\tilde{V} = [v_1, v_2, v_3, v_4]$ be two Trapezoidal fuzzy numbers.

$$\tilde{U} + \tilde{V} = (u_1 + v_1, u_2 + v_2, u_3 + v_3, u_4 + v_4)$$

$$\tilde{U} - \tilde{V} = (u_1 - v_4, u_2 - v_3, u_3 - v_2, u_4 - v_1)$$

$$\tilde{U} \times \tilde{V} = (u_1 v_1, u_2 v_2, u_3 v_3, u_4 v_4)$$

$$\frac{\tilde{U}}{\tilde{V}} = \left(\frac{u_1}{v_4}, \frac{u_2}{v_3}, \frac{u_3}{v_2}, \frac{u_4}{v_1} \right)$$

4) Definition 4 (Yager ranking): If the α cut of any fuzzy number \tilde{A} is $[A_L(\alpha) + A_R(\alpha)]$, then its ranking index $I(\tilde{A})$ is

said to be is $\frac{1}{2} \int_0^1 [A_L(\alpha) + A_R(\alpha)] d\alpha$.

5) Definition 5 (Alpha cut): The set of elements that belong to the fuzzy set \tilde{A} atleast to the degree α is called the α level set or α cut. $A(\alpha) = x \in X : \mu_{\tilde{A}}(x) \geq \alpha$

6) Definition 6 (beta distribution): The trapezoidal fuzzy number $\tilde{B}_M = (m_1, m_2, m_3, m_4)$ are defuzzified by using the following beta distribution formula,

$$\eta_{\tilde{B}_M} = \frac{2m_1 + 7m_2 + 7m_3 + 2m_4}{18}$$

III. NOTATIONS

A. Crisp Parameters

S_{Ja} = Setup cost per cycle

S_{Jb} = Production cost per unit

S_{Jc} = Inventory cost per unit product in a time unit

S_{Jd} = Waste disposal fixed cost per cycle

- S_{Jk} = Average production emission cost per unit
- S_{Jk} = Average inventory emission cost per unit
- S_{Jdk} = Average waste disposal emission cost per unit
- S_{Jek} = Average Recycling emission cost per unit
- K_M = Production rate
- T_M = Demand rate
- R_q = Recycling rate
- R_j = Recycling Cost
- A_{epc} = Average production energy consumption per unit
- A_{erc} = Average recycling energy consumption per unit
- A_{cws} = Average energy consumption per warehouse space unit
- u_s = Space occupied by a unit product
- A_{sw} = Average weight of solid waste produced per unit product
- E_{Gse} = Energy generated standard emission
- E_{Dsw} = Disposal standard emission per ton of solid waste
- X_{tax} = Carbon price or tax
- S_c = Screening cost
- G_{ct} = Green technology investment cost
- Z_i = Proportion of the return of the demand
- J_i = Social costs of vehicle emissions
- W_p = Speed Average
- x_i = Fixed cost for each trip
- y_i = Variable cost
- f = Mileage
- J_{XT} = optimum order size
- $TC(J_{XT})$ = Total inventory cost

IV. FUZZY PARAMETERS

- \tilde{S}_{Ja} = Fuzzy Setup cost per cycle
- \tilde{S}_{Jb} = Fuzzy Production cost per unit
- \tilde{S}_{Jc} = Fuzzy Inventory cost per unit product in a time unit
- \tilde{S}_{Jd} = Fuzzy Waste disposal fixed cost per cycle
- S_{Jk} = Fuzzy average production emission cost per unit
- \tilde{S}_{Jk} = Fuzzy Average inventory emission cost per unit
- \tilde{S}_{Jdk} = Average waste disposal emission cost per unit

\tilde{S}_{Jek} = Fuzzy Average Recycling emission cost per unit

\tilde{R}_q = Fuzzy Recycling rate

\tilde{R}_j = Fuzzy Recycling Cost

\tilde{S}_c = Fuzzy screening cost

\tilde{G}_{ct} = Fuzzy green technology investment cost

\tilde{x}_i = Fuzzy cost for each trip

\tilde{J}_i = Fuzzy social costs of vehicle emissions

Assumptions

1. Demand is taken as known and constant.
2. uncertain parameters are taken as trapezoidal fuzzy numbers.

V. MATHEMATICAL MODEL

A. Mathematical Formulation in Crisp Sense

In this model, carbon emissions occur in the stages of production, shipping, storage, disposal of waste, and recycling. Carbon taxation is imposed on each unit of carbon released. Items that can be recycled are taken to the recycling facility, while those that cannot be recycled are go to the disposal process. To minimize the damaging effect of carbon emissions, investment in green technology is considered.

B. Crisp Sense

Total inventory cost can be calculated by adding the setup cost, production cost, inventory holding cost, waste disposal cost, recycling cost, transportation cost, screening cost and green technology cost

Setup expenses are given as, $\left[\frac{S_{Ja} T_M}{J_{XT}} \right]$

Production expenses are given as, $\left[(S_{Jb} + S_{Jbk}) T_M \right]$, where $S_{Jbk} = (A_{epc} \times E_{Gse} \times X_{tax})$

Inventory holding expenses are given as, $\left[\frac{(S_{Jc} + S_{Jck})(K_M - T_M) J_{XT}}{2K_M} \right]$, where $S_{Jck} = (u_s \times A_{cws} \times E_{Gse} \times X_{tax})$

Expenses for waste disposal are given by, $\left[\frac{S_{Jd} T_M}{J_{XT}} + S_{Jdk} T_M \right]$, where $S_{Jdk} = (A_{sw} \times E_{Dsw} \times X_{tax})$

Expenses for recycling are given as, $\left[\frac{(R_q \times R_j) T_M}{J_{XT}} + \frac{S_{Jek} T_M}{J_{XT}} \right]$, where $S_{Jek} = (A_{erc} \times E_{Gse} \times X_{tax})$

Transportation cost is given as, $\left[\frac{2x_i T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2J_i f T_M}{J_{XT}} \right]$

Screening cost is expressed as, $\left[\frac{S_c T_M}{J_{XT}} \right]$

Green technology cost is given as, $\left[\frac{G_{ct} T_M}{J_{XT}} \right]$

Total inventory cost

$$TC(J_{XT}) = \left[\begin{aligned} & \frac{S_{Ja} T_M}{J_{XT}} + S_{Jb} T_M + S_{Jbk} T_M + \frac{S_{Jc} (K_M - T_M)}{2K_M} + \frac{S_{Jck} (K_M - T_M) J_{XT}}{2K_M} + \frac{S_{Jd} T_M}{J_{XT}} \\ & + S_{Jdk} T_M + \frac{(R_q \times R_j) T_M}{J_{XT}} + \frac{S_{Jek} T_M}{J_{XT}} + \frac{2x_i T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2J_i f T_M}{J_{XT}} + \frac{s_c T_M}{J_{XT}} + \frac{G_{ct} T_M}{J_{XT}} \end{aligned} \right] \dots (1)$$

By differentiating the equation (1) with respect to J_{XT} and equating to 0. We get optimal order quantity,

$$J_{XT}^* = \sqrt{\frac{2T_M \left[\frac{S_{Ja} + S_{Jd} + (R_q \times R_j)}{w_p} + S_{Jek} + 2x_i + S_c + G_{ct} \right]}{[S_{Jc} + S_{Jck}] \left(1 - \frac{T_M}{K_M} \right)}} \dots (2)$$

C. Fuzzy sense

$$T\tilde{C}(J_{XT}) = \left[\begin{aligned} & \frac{\tilde{S}_{Ja} T_M}{J_{XT}} + \tilde{S}_{Jb} T_M + \tilde{S}_{Jbk} T_M + \frac{\tilde{S}_{Jc} (K_M - T_M)}{2K_M} + \frac{\tilde{S}_{Jck} (K_M - T_M) J_{XT}}{2K_M} + \frac{\tilde{S}_{Jd} T_M}{J_{XT}} \\ & + \tilde{S}_{Jdk} T_M + \frac{(\tilde{R}_q \times \tilde{R}_j) T_M}{J_{XT}} + \frac{\tilde{S}_{Jek} T_M}{J_{XT}} + \frac{2\tilde{x}_i T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2\tilde{J}_i f T_M}{J_{XT}} + \frac{\tilde{s}_c T_M}{J_{XT}} + \frac{\tilde{G}_{ct} T_M}{J_{XT}} \end{aligned} \right] \dots (3)$$

Parameters are taken as trapezoidal fuzzy numbers.

$$\begin{aligned} \tilde{S}_{Ja} &= (S_{Ja_1}, S_{Ja_2}, S_{Ja_3}, S_{Ja_4}), \tilde{S}_{Jb} = (S_{Jb_1}, S_{Jb_2}, S_{Jb_3}, S_{Jb_4}), \tilde{S}_{Jc} = (S_{Jc_1}, S_{Jc_2}, S_{Jc_3}, S_{Jc_4}) \\ \tilde{S}_{Jd} &= (S_{Jd_1}, S_{Jd_2}, S_{Jd_3}, S_{Jd_4}), \tilde{S}_{Jbk} = (S_{Jbk_1}, S_{Jbk_2}, S_{Jbk_3}, S_{Jbk_4}), \tilde{S}_{Jck} = (S_{Jck_1}, S_{Jck_2}, S_{Jck_3}, S_{Jck_4}), \\ \tilde{S}_{Jdk} &= (S_{Jdk_1}, S_{Jdk_2}, S_{Jdk_3}, S_{Jdk_4}), \tilde{S}_{Jek} = (S_{Jek_1}, S_{Jek_2}, S_{Jek_3}, S_{Jek_4}), \tilde{R}_q = (R_{q_1}, R_{q_2}, R_{q_3}, R_{q_4}), \\ \tilde{R}_j &= (R_{j_1}, R_{j_2}, R_{j_3}, R_{j_4}) \\ \tilde{x}_i &= (x_{i_1}, x_{i_2}, x_{i_3}, x_{i_4}), \tilde{J}_i = (J_{i_1}, J_{i_2}, J_{i_3}, J_{i_4}), \tilde{S}_c = (s_{c_1}, s_{c_2}, s_{c_3}, s_{c_4}), \tilde{G}_{ct} = (G_{ct_1}, G_{ct_2}, G_{ct_3}, G_{ct_4}) \end{aligned}$$

D. Yager ranking Approach

Let $S_{Ja}, S_{Jb}, S_{Jc}, S_{Jd}, S_{Jbk}, S_{Jck}, S_{Jdk}, R_q, R_j, x_i, J_i, S_{Jek}, S_c, G_{ct}$ be

$$S_{Ja}(\alpha_{S_{Ja}}) = \left[L^{-1}_{S_{Ja}}(\alpha_{S_{Ja}}), R^{-1}_{S_{Ja}}(\alpha_{S_{Ja}}) \right]$$

$$S_{Jb}(\alpha_{S_{Jb}}) = \left[L^{-1}_{S_{Jb}}(\alpha_{S_{Jb}}), R^{-1}_{S_{Jb}}(\alpha_{S_{Jb}}) \right]$$

$$S_{Jc}(\alpha_{S_{Jc}}) = \left[L^{-1}_{S_{Jc}}(\alpha_{S_{Jc}}), R^{-1}_{S_{Jc}}(\alpha_{S_{Jc}}) \right]$$

$$S_{Jd}(\alpha_{S_{Jd}}) = \left[L^{-1}_{S_{Jd}}(\alpha_{S_{Jd}}), R^{-1}_{S_{Jd}}(\alpha_{S_{Jd}}) \right]$$

$$S_{Jbk}(\alpha_{S_{Jbk}}) = \left[L^{-1}_{S_{Jbk}}(\alpha_{S_{Jbk}}), R^{-1}_{S_{Jbk}}(\alpha_{S_{Jbk}}) \right]$$

$$S_{Jck}(\alpha_{S_{Jck}}) = \left[L^{-1}_{S_{Jck}}(\alpha_{S_{Jck}}), R^{-1}_{S_{Jck}}(\alpha_{S_{Jck}}) \right]$$

$$S_{Jdk}(\alpha_{S_{Jdk}}) = \left[L^{-1}_{S_{Jdk}}(\alpha_{S_{Jdk}}), R^{-1}_{S_{Jdk}}(\alpha_{S_{Jdk}}) \right]$$

$$S_{Jek}(\alpha_{S_{Jek}}) = \left[L^{-1}_{S_{Jek}}(\alpha_{S_{Jek}}), R^{-1}_{S_{Jek}}(\alpha_{S_{Jek}}) \right]$$

$$R_q(\alpha_{R_q}) = \left[L^{-1}_{R_q}(\alpha_{R_q}), R^{-1}_{R_q}(\alpha_{R_q}) \right]$$

$$R_j(\alpha_{R_j}) = \left[L^{-1}_{R_j}(\alpha_{R_j}), R^{-1}_{R_j}(\alpha_{R_j}) \right]$$

$$x_i(\alpha_{x_i}) = \left[L^{-1}_{x_i}(\alpha_{x_i}), R^{-1}_{x_i}(\alpha_{x_i}) \right]$$

$$J_i(\alpha_{J_i}) = \left[L^{-1}_{J_i}(\alpha_{J_i}), R^{-1}_{J_i}(\alpha_{J_i}) \right]$$

$$S_c(\alpha_{S_c}) = \left[L^{-1}_{S_c}(\alpha_{S_c}), R^{-1}_{S_c}(\alpha_{S_c}) \right]$$

$$G_{ct}(\alpha_{G_{ct}}) = \left[L^{-1}_{G_{ct}}(\alpha_{G_{ct}}), R^{-1}_{G_{ct}}(\alpha_{G_{ct}}) \right]$$

$$K_1(\alpha_{S_{Ja}}) = \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{Ja}}(\alpha_{S_{Ja}}) d\alpha_{S_{Ja}} + \int_0^1 R^{-1}_{S_{Ja}}(\alpha_{S_{Ja}}) d\alpha_{S_{Ja}} \right]$$

$$K_2(\alpha_{S_{Jb}}) = \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{Jb}}(\alpha_{S_{Jb}}) d\alpha_{S_{Jb}} + \int_0^1 R^{-1}_{S_{Jb}}(\alpha_{S_{Jb}}) d\alpha_{S_{Jb}} \right]$$

$$K_3(\alpha_{S_{Jc}}) = \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{Jc}}(\alpha_{S_{Jc}}) d\alpha_{S_{Jc}} + \int_0^1 R^{-1}_{S_{Jc}}(\alpha_{S_{Jc}}) d\alpha_{S_{Jc}} \right]$$

$$K_4(\alpha_{S_{Jd}}) = \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{Jd}}(\alpha_{S_{Jd}}) d\alpha_{S_{Jd}} + \int_0^1 R^{-1}_{S_{Jd}}(\alpha_{S_{Jd}}) d\alpha_{S_{Jd}} \right]$$

$$\begin{aligned}
 K_5(\alpha_{R_q}, \alpha_{R_j}) &= \frac{1}{4} \left[\int_0^1 L^{-1}_{R_q}(\alpha_{R_q}) d\alpha_{R_q} \cdot \int_0^1 L^{-1}_{\alpha_{R_j}}(\alpha_{\alpha_{R_j}}) d\alpha_{R_j} + \int_0^1 R^{-1}_{R_q}(\alpha_{R_q}) d\alpha_{R_q} \cdot \int_0^1 R^{-1}_{\alpha_{R_j}}(\alpha_{\alpha_{R_j}}) d\alpha_{R_j} \right] \\
 K_6(\alpha_{S_{j b k}}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{j b k}}(\alpha_{S_{j b k}}) d\alpha_{S_{j b k}} + \int_0^1 R^{-1}_{S_{j b k}}(\alpha_{S_{j b k}}) d\alpha_{S_{j b k}} \right] \\
 K_7(\alpha_{S_{j c k}}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{j c k}}(\alpha_{S_{j c k}}) d\alpha_{S_{j c k}} + \int_0^1 R^{-1}_{S_{j c k}}(\alpha_{S_{j c k}}) d\alpha_{S_{j c k}} \right] \\
 K_8(\alpha_{S_{j d k}}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{j d k}}(\alpha_{S_{j d k}}) d\alpha_{S_{j d k}} + \int_0^1 R^{-1}_{S_{j d k}}(\alpha_{S_{j d k}}) d\alpha_{S_{j d k}} \right] \\
 K_9(\alpha_{S_{j e k}}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{S_{j e k}}(\alpha_{S_{j e k}}) d\alpha_{S_{j e k}} + \int_0^1 R^{-1}_{S_{j e k}}(\alpha_{S_{j e k}}) d\alpha_{S_{j e k}} \right] \\
 K_{10}(\alpha_{x_i}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{x_i}(\alpha_{x_i}) d\alpha_{x_i} + \int_0^1 R^{-1}_{x_i}(\alpha_{x_i}) d\alpha_{x_i} \right] \\
 K_{11}(\alpha_{J_i}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{J_i}(\alpha_{J_i}) d\alpha_{J_i} + \int_0^1 R^{-1}_{J_i}(\alpha_{J_i}) d\alpha_{J_i} \right] \\
 K_{12}(\alpha_{s_c}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{s_c}(\alpha_{s_c}) d\alpha_{s_c} + \int_0^1 R^{-1}_{s_c}(\alpha_{s_c}) d\alpha_{s_c} \right] \\
 K_{13}(\alpha_{G_{c t}}) &= \frac{1}{2} \left[\int_0^1 L^{-1}_{G_{c t}}(\alpha_{G_{c t}}) d\alpha_{G_{c t}} + \int_0^1 R^{-1}_{G_{c t}}(\alpha_{G_{c t}}) d\alpha_{G_{c t}} \right] \\
 T\tilde{C}(J_{XT}) &= \left[\begin{aligned}
 &\frac{K_1(\alpha_{S_{j a}})T_M}{J_{XT}} + K_2(\alpha_{S_{j b}})T_M + K_6(\alpha_{S_{j b k}})T_M + \frac{K_3(\alpha_{S_{j c}})(K_M - T_M)}{2K_M} \\
 &+ \frac{K_7(\alpha_{S_{j c k}})(K_M - T_M)J_{XT}}{2K_M} + \frac{K_4(\alpha_{S_{j d}})K_M}{J_{XT}} \\
 &+ K_8(\alpha_{S_{j d k}})T_M + \frac{K_5(\alpha_{R_q}, \alpha_{R_j})T_M}{J_{XT}} + \frac{K_9(\alpha_{S_{j e k}})T_M}{J_{XT}} + 2\frac{K_{10}(\alpha_{x_i})T_M}{J_{XT}} \\
 &+ y_i f T_M + 1 + z_i + \frac{2K_{11}(\alpha_{J_i})f T_M}{J_{XT}} + \frac{K_{12}(\alpha_{s_c})T_M}{J_{XT}} + \frac{K_{13}(\alpha_{G_{c t}})T_M}{J_{XT}}
 \end{aligned} \right] \dots (4)
 \end{aligned}$$

Differentiating equation (4) with respect to J_{XT} and equating to 0, we get fuzzy optimal order quantity in yager ranking method,

$$\tilde{J}_{XT}^* = \sqrt[2T_M]{\frac{\left[K_1(\alpha_{S_{Ja}}) + K_4(\alpha_{S_{Jd}}) + K_5(\alpha_{R_q}, \alpha_{R_j}) + K_9(\alpha_{S_{Jek}}) + 2K_{10}(\alpha_{x_i}) + K_{12}(\alpha_{s_c}) + K_{13}(\alpha_{G_{ct}}) + \frac{2K_{11}(\alpha_{J_i})fT_M}{w_p} \right]}{\left[K_3(\alpha_{S_{Jc}}) + K_7(\alpha_{S_{Jck}}) \right] \left(1 - \frac{T_M}{K_M} \right)}} \dots\dots\dots (5)$$

E. Beta Distribution Method

By using the beta distribution method formula, $\eta_{\tilde{c}} = \frac{2m_1 + 7m_2 + 7m_3 + 2m_4}{18}$, we get fuzzy total inventory cost

$$T\tilde{C}(J_{XT}) = \left[\begin{aligned} &2 \left(\frac{S_{Ja_1} T_M}{J_{XT}} + (S_{Jb_1} + S_{Jbk_1}) T_M + \frac{(S_{Jc_1} + S_{Jck_1})(K_M - T_M)}{2K_M} + \frac{S_{Jd_1} K_M}{J_{XT}} + S_{Jdk_1} T_M \right. \\ &\quad \left. + \frac{(R_{q_1} \times R_{j_1}) T_M}{J_{XT}} + \frac{S_{Jek_1} T_M}{J_{XT}} + \frac{2x_{i_1} T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2J_{i_1} f T_M}{J_{XT}} + \frac{S_{c_1} T_M}{J_{XT}} + \frac{G_{ct_1} T_M}{J_{XT}} \right) \\ &+7 \left(\frac{S_{Ja_2} T_M}{J_{XT}} + (S_{Jb_2} + S_{Jbk_2}) T_M + \frac{(S_{Jc_2} + S_{Jck_2})(K_M - T_M)}{2K_M} + \frac{S_{Jd_2} K_M}{J_{XT}} + S_{Jdk_2} T_M \right. \\ &\quad \left. + \frac{(R_{q_2} \times R_{j_2}) T_M}{J_{XT}} + \frac{S_{Jek_2} T_M}{J_{XT}} + \frac{2x_{i_2} T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2J_{i_2} f T_M}{J_{XT}} + \frac{S_{c_2} T_M}{J_{XT}} + \frac{G_{ct_2} T_M}{J_{XT}} \right) \\ &+7 \left(\frac{S_{Ja_3} T_M}{J_{XT}} + (S_{Jb_3} + S_{Jbk_3}) T_M + \frac{(S_{Jc_3} + S_{Jck_3})(K_M - T_M)}{2K_M} + \frac{S_{Jd_3} K_M}{J_{XT}} + S_{Jdk_3} T_M \right. \\ &\quad \left. + \frac{(R_{q_3} \times R_{j_3}) T_M}{J_{XT}} + \frac{S_{Jek_3} T_M}{J_{XT}} + \frac{2x_{i_3} T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2J_{i_3} f T_M}{J_{XT}} + \frac{S_{c_3} T_M}{J_{XT}} + \frac{G_{ct_3} T_M}{J_{XT}} \right) \\ &+2 \left(\frac{S_{Ja_4} T_M}{J_{XT}} + (S_{Jb_4} + S_{Jbk_4}) T_M + \frac{(S_{Jc_4} + S_{Jck_4})(K_M - T_M)}{2K_M} + \frac{S_{Jd_4} K_M}{J_{XT}} + S_{Jdk_4} T_M \right. \\ &\quad \left. + \frac{(R_{q_4} \times R_{j_4}) T_M}{J_{XT}} + \frac{S_{Jek_4} T_M}{J_{XT}} + \frac{2x_{i_4} T_M}{J_{XT}} + y_i f T_M + 1 + z_i + \frac{2J_{i_4} f T_M}{J_{XT}} + \frac{S_{c_4} T_M}{J_{XT}} + \frac{G_{ct_4} T_M}{J_{XT}} \right) \end{aligned} \right] \dots\dots (6)$$

Differentiating equation (6) with respect to J_{XT} and equating to 0. We get fuzzy optimal order quantity using beta distribution formula.

$$\tilde{J}_{XT}^* = \frac{\left[\begin{aligned} & \left(2S_{Ja_1} + 7S_{Ja_2} + 7S_{Ja_3} + 2S_{Ja_4} \right) + \left(2S_{Jd_1} + 7S_{Jd_2} + 7S_{Jd_3} + 7S_{Jd_4} \right) + \\ & \left(2(R_{q_1} \times R_{j_1}) + 7(R_{q_2} \times R_{j_2}) + 7(R_{q_3} \times R_{j_3}) + 2(R_{q_4} \times R_{j_4}) \right) + \\ & 2T_M \left(\left(2S_{Jek_1} + 7S_{Jek_2} + 7S_{Jek_3} + 2S_{Jek_4} \right) \right) + 2(2x_{i_1} + 7x_{i_2} + 7x_{i_3} + 2x_{i_4}) \\ & \left(2S_{c_1} + 7S_{c_2} + 7S_{c_3} + 2S_{c_4} \right) + \left(2G_{ct_1} + 7G_{ct_2} + 7G_{ct_3} + 2G_{ct_4} \right) + \\ & \left(\frac{2f}{W_p} (2J_{i_1} + 7J_{i_2} + 7J_{i_3} + 2J_{i_4}) \right) \end{aligned} \right]}{\left[\left(2S_{Jc_1} + 7S_{Jc_2} + 7S_{Jc_3} + 2S_{Jc_4} \right) + \left(2S_{Jck_1} + 7S_{Jck_2} + 7S_{Jck_3} + 2S_{Jck_4} \right) \right] \left(1 - \frac{T_M}{K_M} \right)} \dots (7)$$

VI. NUMERICAL EXAMPLE

A. Crisp Data Values

$T_M = 140, K_M = 200, S_{Ja} = 25, S_{Jb} = 9, S_{Jc} = 4.5, S_{Jd} = 8, X_{tax} = 130, R_q = 50, R_j = 120, A_{epc} = 50, A_{erc} = 42.30,$
 $A_{cws} = 5, u_s = 1.7, E_{Dsw} = 0.3, A_{sw} = 10, E_{Gse} = 0.5, S_c = 0.25, G_{ct} = 300, Z_i = 0.75, J_i = 120, W_p = 80, x_i = 60,$
 $y_i = 35, f = 170$

$$S_{Jbk} = \left(50 \times \left(\frac{0.5}{1000} \right) \times 130 \right) = 3.25, S_{Jck} = \left(1.7 \times 5 \times \left(\frac{0.5}{1000} \right) \times 130 \right) = 0.5525,$$

$$S_{Jdk} = \left(\left(\frac{10}{1000} \right) \times 0.3 \times 130 \right) = 0.39, S_{Jek} = \left(42.30 \times \left(\frac{0.5}{1000} \right) \times 130 \right) = 2.75$$

B. Fuzzy data values

$$\tilde{S}_{Ja} = (23, 24, 26, 27), \tilde{S}_{Jb} = (7, 8, 10, 11), \tilde{S}_{Jc} = (4.3, 4.4, 4.6, 4.7), \tilde{S}_{Jd} = (6, 7, 9, 10), \tilde{x}_i = (58, 59, 61, 62)$$

$$\tilde{R}_q = (118, 119, 121, 122), \tilde{R}_j = (48, 49, 51, 52), \tilde{s}_c = (0.23, 0.24, 0.26, 0.27), \tilde{G}_{ct} = (298, 299, 301, 302),$$

$$\tilde{J}_i = (118, 119, 121, 122), \tilde{S}_{Jbk} = (3.248, 3.249, 3.251, 3.252), \tilde{S}_{Jdk} = (0.37, 0.38, 0.41, 0.42), \tilde{S}_{Jek} = (2.73, 2.74, 2.76, 2.77),$$

$$\tilde{S}_{Jck} = (0.5523, 0.5524, 0.5526, 0.5527)$$

Solution in crisp model

Using the above data in the equation (2) we obtain the optimal order quantity as $v = 80.21$. Using the equation (1) we obtain the total inventory cost as $TC = Rs. 846990.69$

Solution in fuzzy model

Yager ranking solution

Using the above data in the equation (5) we obtain the optimal order quantity as $v^* = 60.53$. Using the equation (4) we obtain the total inventory cost as $TC^* = Rs. 843992.778$

Beta distribution solution

Using the above data in the equation (7) we obtain the optimal order quantity as $v^* = 80.22$. Using the equation (6) we obtain the total inventory cost as $TC^* = Rs. 846992.768$

VII. CONCLUSION

Recycling is indispensable for the advancement of environmental preservation in a number of ways. On an economic scale, it lowers the systematic consumption of essential resources, ultimately resulting in cost savings. It has an advantageous ecological influence through reducing CO₂ emissions and solid waste generation. This paper developed economic manufacturing model with ecological concerns of lowering emissions by adopting recycling and green technology investment in inventory functions. The findings of the research suggest that the Yager ranking method provides a lower solution results than beta distribution method.

REFERENCES

- [1] Poswal P, Chauhan A, Aarya DD, Boadh R, Rajoria YK, Gaiola SU. Optimal strategy for remanufacturing system of sustainable products with trade credit under uncertain scenario. *Materials Today: Proceedings*. 2022;69(2):165-173.
- [2] Karim R, Nakade K. A literature review on the sustainable EPQ model, focusing on carbon emissions and product recycling. *Logistics*. 2022;6(3):1-16.
- [3] Dwicahyani AR, Jauhari WA, Rosyidi CN, Laksono PW. Inventory decisions in a two-echelon system with remanufacturing, carbon emission, and energy effects. *cogent engineering*. 2017;4(1):1-17.
- [4] Daryanto Y, Wee HM. Sustainable economic production quantity models: an approach toward a cleaner production. *Journal of Advanced Management Science* Vol. 2018;6(4): 206-212.
- [5] Priyan S, Mala P, Palanivel M. A cleaner EPQ inventory model involving synchronous and asynchronous rework process with green technology investment. *Cleaner Logistics and Supply Chain*. 2022 Jul 1;4: 100056.
- [6] Shah NH, Patel DG, Shah DB, Prajapati NM. A sustainable production inventory model with green technology investment for perishable products. *Decision Analytics Journal*. 2023 Sep 1;8: 100309.
- [7] Singh R, Mishra VK. Machine learning based fuzzy inventory model for imperfect deteriorating products with demand forecast and partial backlogging under green investment technology. *Journal of the Operational Research Society*. 2024 Jul 2;75(7):1223-38.
- [8] Kumar BA, Paikray SK. Cost optimization inventory model for deteriorating items with trapezoidal demand rate under completely backlogged shortages in crisp and fuzzy environment. *RAIRO-Operations Research*. 2022;56(3):1969-94.
- [9] Maity AK, Maity K, Maiti M. A production–recycling–inventory system with imprecise holding costs. *Applied Mathematical Modelling*. 2008;32(11):2241-53.
- [10] Shekarian E, Olugu EU, Abdul-Rashid SH, Bottani E. A fuzzy reverse logistics inventory system integrating economic order/production quantity models. *International Journal of Fuzzy Systems*. 2016; 18:1141-61.



10.22214/IJRASET



45.98



IMPACT FACTOR:
7.129



IMPACT FACTOR:
7.429



INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Call : 08813907089  (24*7 Support on Whatsapp)