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Post-harvest losses in Indian maize amid increasing food insecurity Analysis using TOPSIS method

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Abstract

Post-harvest loss in harvesting crops Food supply chain from production to their consumption All include loss of food. Spoilage, loss of quality, loss of nutrients, seed By loss of credibility, and loss of business Losses can be classified as weight loss. Importantly, Rotting, mechanical damage, poor handling, temperature, and improper management and handling of moisture Health problems during post-harvest are Major causes of losses. during retailing. "Post-harvest losses" refers measurable in a given production Loss of quantity and quality. These losses can occur at various stages of the post-harvest process. This definition should also take into account product spoilage events. Post-harvest losses of vegetables and fruits From produce in the field to food on a plate Occurs at all points in the value chain. Post-harvest operations include harvesting, Handling, storage, processing, packaging, transportation and include marketing. Diseases Cassava mosaic disease, Cassava brown such as streak and cassava bacterial blight Crop field losses increase and they rot While, large grain borers and cycle moths are the products stored after harvest Major cause of losses. Crop losses per season Vary with the season and from one place to another. Post-harvest Diseases of fruits are caused by pathogenic microorganisms and are a major cause of vegetable losses. Most of the losses are post-harvest Alternaria, Aspergillus, Botrytis, Colletotrichum, Diplodia, Monilinia, Penicillium, Phomopsis, Rhizopus, Mucor and Sclerotinia are caused by fungi such as Alternative: IPR, FP, IPR, and FP. Evaluation Preference: Cultivation, Harvesting, and Drying, Storage, Milling and Paddy transportation. from the result it is seen that Harvesting is got the first rank where as is the Milling is having the lowest rank. The value of the dataset for Post-harvest loss in TOPSIS method shows that it results in Harvesting and top ranking.

Keywords: Cultivation, Harvesting, Drying, Paddy transportation.

Introduction

For harvesting on systematically collected datasets Post management and farmers of various crops, Post-harvest losses, and general social- Economic information and geographic information are included. of farming households, [1] Climate change, like temperature and humidity through changes in various climate variables Post-harvest losses are likely to increase. It also creates an environment for new insect pests to thrive. [2] Global food loss and waste of as much as 32 percent are estimated. Such savings opportunities, global Post-harvest in facing food challenge An urgent need to focus more on loss Recommend. Chronic malnutrition and Such in the face of sluggish yield gains They also question why significant losses persisted wake up [3] Several definitions of post-harvest losses There are, food loss and waste each other An alternative term, PHL is food for human consumption through the stages of the supply chain (FSC). An accidental decrease in the amount of food produced is defined. cause or goal Regardless. [4] Post-harvest losses (PHL) in India Significant and massive economic deficits causing The reduction of PHL in economic, social, and In balancing environmental dimensions Helps achieve stability. Fruits in Indian context and vegetables (F&V) supply chain of PHL To identify important factors [5] Almost 6% storage loss due to inadequate and outdated storage facilities contributes to a A significant proportion of post-harvest losses. In India, about 12 to 16 million tonnes annually of Food grains are wasted, worth about 4 billion US dollars. This amount of food grains To feed about 10% of India's population Adequate and proper storage and storage Reducing losses to meet 10% of India's food will help. requirement. [6] Post-harvest losses and horticulture Deterioration of crop quality are mostly due to pests, Microbial infection, natural ripening processes, and such as heat and drought caused by environmental conditions. and improper Post-harvest handling. [7] Investing In reducing post-harvest losses (PHL) other As with investments, the benefits outweigh the costs carried out if any. Do policy Optimal choices of reporting and mitigation approaches to simplify, levels losses, [8] Investment in reducing post-harvest losses (PHL). So, if the benefits outweigh the costs, other Investments are made in the same way. of losses Quantities of losses at each stage Net of following drivers and mitigation procedures, The principle is to get accurate knowledge of the benefits Inform, optimal choices of mitigation approach Simplification is also important. [9] Post-harvest in food value chains amount (PHL) diet researchers and policy To reduce these losses among classifiers It is increasingly discussed with the design of policies. First, reducing PHL is believed to improve food security because lower PHL will ensure more food availability at lower prices. [10] Despite Comprehensive research and development of remedial techniques, Recommended control measures are limited There are Reducing post-harvest losses There is a lot of information published on related aspects fewer Effects of loss as fully as possible should be analyzed. [11] Evaluated post-harvest and food loss practices A maize farming system in Tanzania. savings Significant losses were incurred

during the phase detected. From the perspective of farmers, climate change, Field damage, and pests main causes of losses [13] The transition Traditional post-harvest operations begin with growing Considerably up to mechanized systems in countries Adds investment costs, especially for machinery, Also engine output, and fuel consumption 6-8 Greenhouse Gas Emissions (GHGE) from leads to increase. However, mechanized Under traditional operations compared to systems Post-harvest losses are significantly higher. [14] Building agro-solids production, post-harvest field processing program will prevent loss reduction in smart agriculture. Many processing programs prevent post-harvest losses in agricultural solids production, apple is susceptible to a limited number of diseases, but it is a large amount of wastage in the post-harvest process. [15]

Materials & Methods

We develop a fuzzy TOPICS analytical Evaluation of customer satisfaction under conditions of uncertainty A model of customer satisfaction system used and general assessment. Validity of assessment Characterization is shown through a case study [1] A model with Prioritize rankings and alternatives Obscure TOPSIS for rendering based approach It is also proposed to provide Throughout this study, Five ATM, POS, Tele Banking, Mobile, and Internet- The authors operationalize a decision-support model of outsourcing for banking services Real banking using data From the case. The proposed model is ITO there to decision makers Helps in decision making It helps to drive the process forward, especially when the parameters are uncertain Where characters are involved and cannot be judged by human judgments. [3] of Fuzzy TOPSIS based approach A model with applications, alternatives Rank and prioritize proposed. Throughout this study, five ATMs, POS, Tele banking, Mobile and Internet-Banking Real Bank to run a decision-support model of outsourcing for services the authors use data from the case. [4] KANUMAN One of the most classical maximization methods, TOPSYS Originally developed by Wang and Lee, this The basic idea of the method is that Selected alternatives positive ideal solution Very short distance and far from negativity Must also have distance. Best solution. [5] Key Between TOPSIS and Fuzzy TOPSIS Difference are evaluation approaches. Qualification Fuzzy TOPSIS Evaluation Performance Defined criteria based on given alternatives is the value of alternatives It is the use of Ambiguous data to disclose. Because given by decision makers Decision-making information is often present inaccurate [6]. A hybrid of the two methods has been highlighted Existing literature. Based on this context, GSD program outcomes as perceived by groups Fuzzy TOPSIS and DEMATEL for estimation A combination of approaches is proposed. [7] A decision AHP and Fuzzy TOPSIS approach and model are proposed. AHP To obtain the weights of the criteria, Obscure TOPSIS is used. Rank ecommerce websites. Weights derived from AHP Fuzzy TOPSIS calculations using them are included in the decision-making process and The rank order is decided based on that these weights. [8] Weight vector of risk factors and ambiguous outcome By using a weighted normal fuzzy The result matrix is generated. Then, FPIS and FNIS From and every failure from FPIS and FNIS The mode distance is calculated respectively. Chen's at the last stage of fuzzy TOPSIS closeness coefficients Processes are received. According to close coefficients, a ranking order of all failure modes is determined. [9] Concepts of FPIS and FNIS are Fuzzy TOPSIS They play an important role practice. A positive ideal solution and the best solution to the negative are the positive or cost criteria, respectively Depending on Maximum or minimum replacement has values. However, in practice, the best a solution is the extreme that can be obtained when evaluating alternatives the values do not need to be the same. [10] To rank Specify alternative and preferred option, far For this purpose, IVSF cosine similarity measure, it's To demonstrate effectiveness and practicality, advertise We use the IVSF set method for strategy selection We proposed. Alternatives and to represent evaluations of criteria. [11] Many based on TOPSIS Criteria for Decision making problems in renewable energy sources A new and A different way has been proposed and some already exist Methods are compared. Then, conflicted Evaluate possible alternatives between criteria To do, related to renewable energy sources A selection of experts is selected. Also, the decision matrix and criterion weights are fuzzy Measured using linguistic values that are converted to values. [12] of To show the efficiency of the weighting method Location is a problem A usage example is provided. Finally, the results of the proposed method, proposed Explanation provided to show validation of interval value Related to the example with some methods from the literature are compared. Reluctance TOPSIS method. [13] This study is a robust ERP selection framework Proposes, including FLPR and alternative ERP systems of evaluation criteria calculated by ranking the weights are obtained by TOPSIS. Food Turkish total real life application in company implemented. [14] AHP is relative weighting of constraints TOPSIS is used for prioritization. RL Developing, developing and investing in technology Doing is the highest ranking for RL adoption are solutions. A constraint of Indian Electronics Industry Beyond empirical evidence supports this framework. RL adoption. [15]

Cultivation: Cultivation means growing something or improving its growth, especially crops. Cultivation can be used more figuratively to refer to the same kind of process, besides being an abstract thing like a business or friendship.

Harvesting: Harvesting is the harvesting of the useful part or parts of the plant Collection is, and all nutrients By the time they are formed and reach edible parts carried out. suitable maturity. Generally, harvesting 10. Grains reach physiological maturity Or after 15 days.

Drying: Drying is a solid, semi-solid, or Evaporation of water or another solvent from a liquid A mass transfer involving removal is the process. This process is often Before selling products or as a final manufacturing step is used packaged.

Milling: Grinding A milling tool cuts the material in a rotating motion is a process. As with drilling, different A variety of diameters and different hard nesses This is possible with a wide array of tools. The plant is moving Since rotating to get a clean finish on the milled hole Speed should be high.

Storage: Storage is where a computer temporarily or helps retain data permanently mechanism. Flash drives and hard disks most storage devices are digital Because they are the basic component of devices Such as videos, documents, images, and raw data Users to protect all types of information allow

Paddy transportation: About 40 million tonnes Food grains by FCI carried throughout year. Movement of food grains is done through railways and waterways. More than 85% of stock movement is done by rail.

Result and discussions

TABLE 1. Post-harvest losses in TOPSIS

	DATA SET			
	IPR(high)	FP(high)	IPR(Low)	FP(Low)
Cultivation	47.08	169.53	19.15	32.05
Harvesting	73.12	172.97	13.69	37.30
Drying	48.08	142.58	15.18	33.10
Storage	51.17	178.28	14.60	37.59
Milling	37.33	136.41	17.96	38.89
Paddy transportation	55.07	166.00	12.55	35.55

These Table 1 TOPSIS of Post-harvest losses Alternative: IPR, FP, IPR, and FP. Evaluation Preference: Cultivation, Harvesting, and Drying, Storage, Milling and Paddy transportation. IPR (high) the Harvesting it is seen that is showing the highest value for Milling is showing the lowest value. FP (high) it is seen that Storage is showing the highest value for Milling is showing the lowest value. IPR (Low) the is seen that Cultivation is showing the highest value for Paddy transportation is showing the lowest value. FP (Low) and it is seen that Hand drill is showing the highest value for corded drill is showing the lowest value.

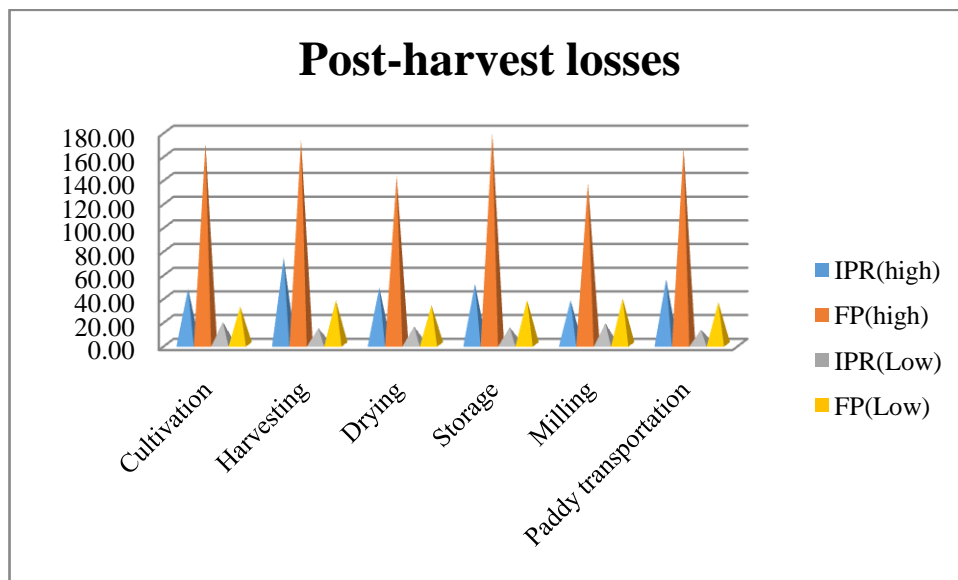


FIGURE 1. Post-harvest losses

These FIGURE 1 TOPSIS of Post-harvest losses Alternative: IPR, FP, IPR, and FP. Evaluation Preference: Cultivation, Harvesting, and Drying, Storage, Milling and Paddy transportation.

TABLE 2. Squire Rote of matrix

2216.526	28740.421	366.7225	1027.203
5346.534	29918.621	187.4161	1391.29
2311.686	20329.056	230.4324	1095.61
2618.369	31783.758	213.16	1413.008
1393.529	18607.688	322.5616	1512.432
3032.705	27556	157.5025	1263.803

Table 2 shows the Squire Rote of matrix value.

TABLE 3. Post-harvest losses in Normalized Data

Normalized Data			
IPR(high)	FP(high)	IPR(Low)	FP(Low)
0.4	1.439	0.527	0.399
0.62	1.468	0.377	0.465
0.408	1.21	0.418	0.412
0.434	1.513	0.402	0.468
0.317	1.158	0.494	0.485
0.467	1.409	0.345	0.443

Table 3 Normalized Data shows the informational set for the Cultivation, Harvesting, and Drying, Storage, Milling and Paddy transportation. The Normalized data is calculated from the data set value is divided by the sum of the square root of the column value.

TABLE 4. Weight

Weight			
0.25	0.25	0.25	0.25
0.25	0.25	0.25	0.25
0.25	0.25	0.25	0.25
0.25	0.25	0.25	0.25
0.25	0.25	0.25	0.25
0.25	0.25	0.25	0.25

Table 4 Weight shows the informational set for the weight all same value 0.25.

TABLE 5. Post-harvest losses in Weighted normalized result matrix

Weighted normalized decision matrix			
0.1	0.36	0.132	0.1
0.155	0.367	0.094	0.116
0.102	0.302	0.104	0.103
0.109	0.378	0.1	0.117
0.079	0.289	0.124	0.121
0.117	0.352	0.086	0.111

Table 3 Normalized Data shows the informational set for the Cultivation, Harvesting, and Drying, Storage, Milling and Paddy transportation. The Normalized data is calculated from the data set value is divided by the sum of the square root of the column value.

TABLE 6. Post-harvest losses in Positive Matrix

Positive Matrix			
0.155	0.378	0.086	0.1
0.155	0.378	0.086	0.1
0.155	0.378	0.086	0.1
0.155	0.378	0.086	0.1
0.155	0.378	0.086	0.1
0.155	0.378	0.086	0.1

Table 6 Positive Matrix shows the informational set for the value IPR (high) 0.155, FP (high) 0.378, IPR (Low) 0.086, FP (Low) 0.1097.

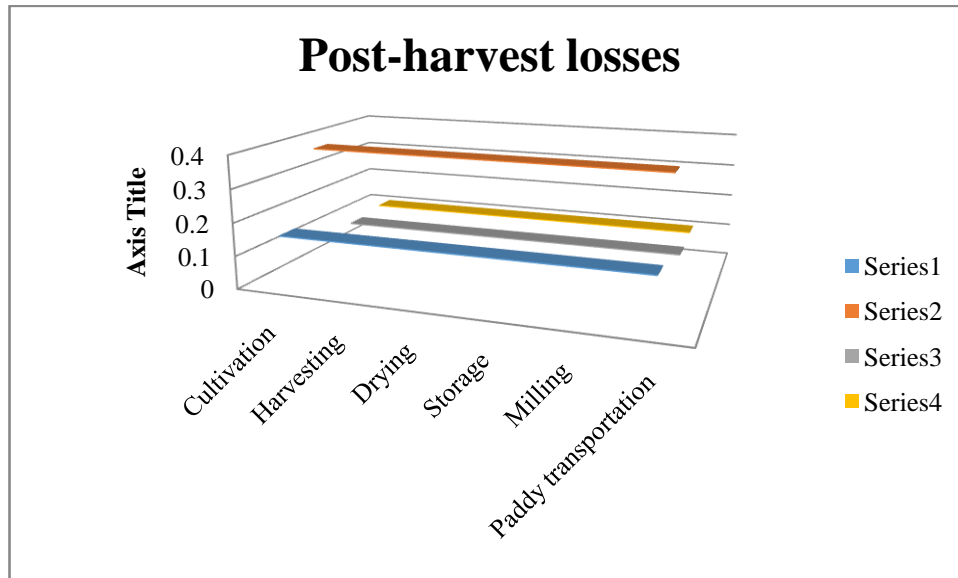


FIGURE 2. Positive Matrix

Figure 2 Positive Matrix shows the informational set for the value IPR (high) 0.155, FP (high) 0.378, IPR (Low) 0.086, FP (Low) 0.1097.

TABLE 7. Post-harvest losses in Negative matrix

	Negative matrix			
Cultivation	0.079	0.289	0.132	0.121158
Harvesting	0.079	0.289	0.132	0.121158
Drying	0.079	0.289	0.132	0.121158
Storage	0.079	0.289	0.132	0.121158
Milling	0.079	0.289	0.132	0.121158
Paddy transportation	0.079	0.289	0.132	0.121158

Table 7 Negative matrix shows the informational set for the value IPR (high) 0.079, FP (high) 0.289, IPR (Low) 0.132, FP (Low) 0.121158

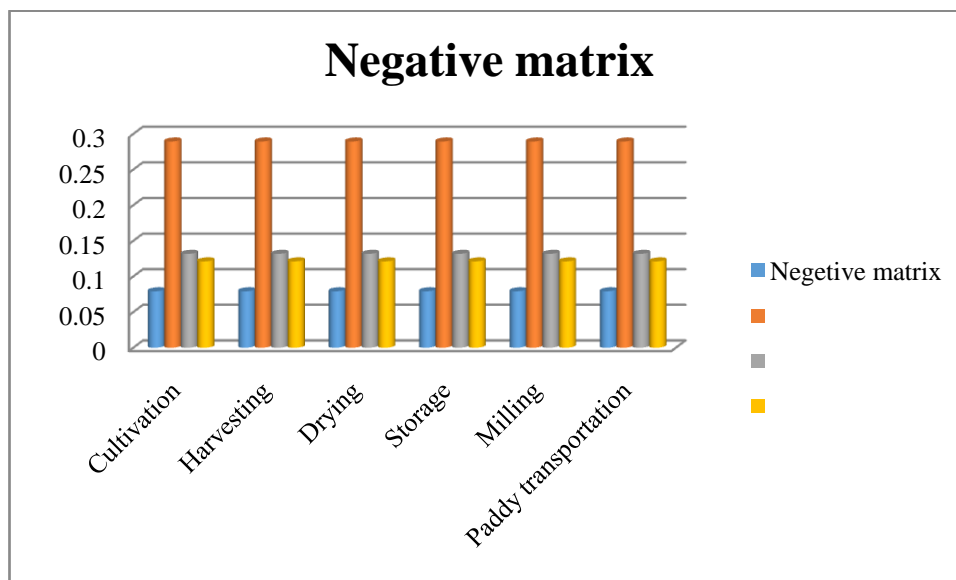


FIGURE 3. Negative matrix

Figure 3 Negative matrix shows the informational set for the value IPR (high) 0.079, FP (high) 0.289, IPR (Low) 0.132, FP (Low) 0.121158

TABLE 8. Post-harvest losses in Si Positive & Si Negative & Ci

	SI Plus	SI Plus	Ci
Cultivation	0.073882	0.0738816	0.507993
Harvesting	0.021353	0.0213526	0.84336
Drying	0.09432	0.0943198	0.308033
Storage	0.051626	0.0516264	0.656653
Milling	0.124478	0.1244784	0.061715
Paddy transportation	0.047581	0.0475809	0.645821

Table 8 Si Positive & Si Negative & Ci shows the graphical representation

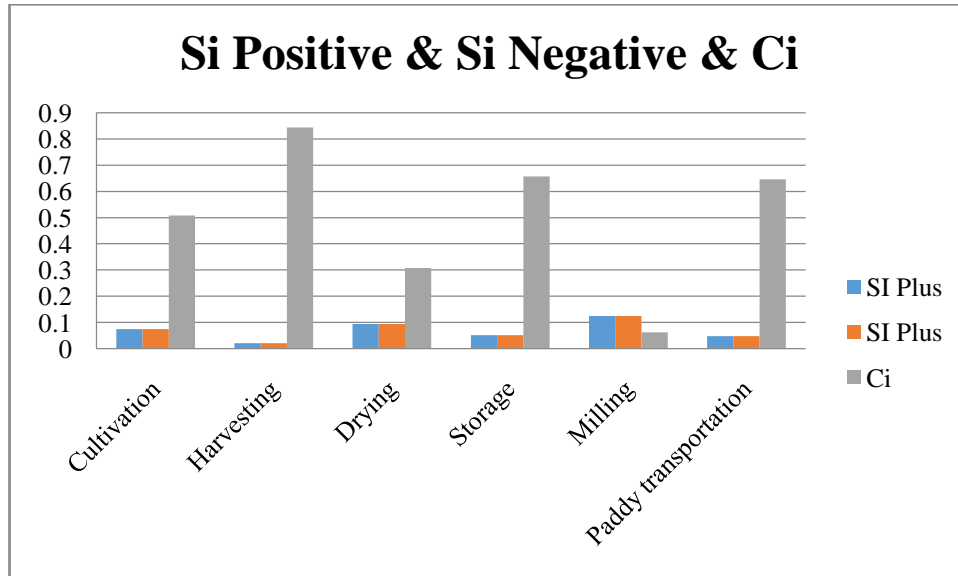


FIGURE 4. Post-harvest losses in Si Positive & Si Negative & Ci

Figure 4 Si Positive & Si Negative & Ci shows the graphical representation

TABLE 9. Post-harvest losses in Rank

	Rank
Cultivation	4
Harvesting	1
Drying	5
Storage	2
Milling	6
Paddy transportation	3

Table 9 shows the from the result it is seen that Harvesting is got the first rank where as is the Milling is having the lowest rank.

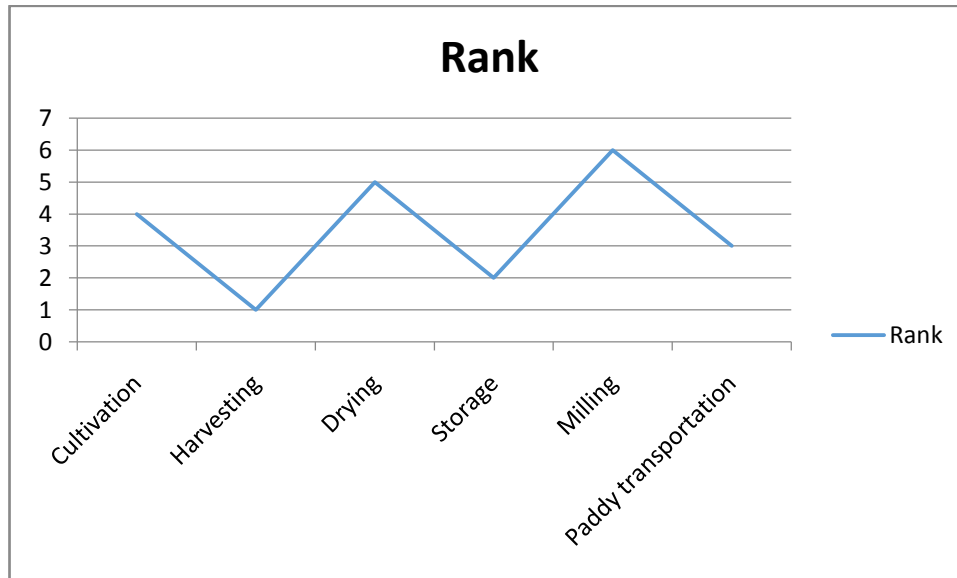


FIGURE 5. Rank

Figure 5 shows the from the result it is seen that Harvesting is got the first rank where as is the Milling is having the lowest rank.

Conclusion

From the result it is seen that Harvesting is got the first rank where as is the Milling is having the lowest rank. Significant Maize during post-harvest handling Loss in both quantity and quality of grains occurs. The reasons for this are biological, socio- Economics and policy related. For these identify the causative biological factors and harvest Research and private in post-technology development appropriate government for both sector involvements Address these by formulating policies can do propagation. of infrastructure such as storage facilities in developing countries Construction, knowledge from industrialized countries, and For reduction of PHL, with the transfer of technologies led to Proper policy implementation and appropriate Using techniques/tools/techniques The need to remove or reduce PHL by has Identify important causal factors and understand their position in the hierarchy Their mathematical modeling is used by decision makers To delete or reducing them. Hence, there is an urgent need to intervene in the existing problems. Additionally, Effective and efficient intervention Policies and strategies should be developed. Many based on TOPSIS Criteria for Decision making problems in renewable energy sources a new and A different way has been proposed and some already exist Methods are compared. Then, conflicted Evaluate possible alternatives between criteria To do, related to renewable energy sources A selection of experts is selected. Also, the decision matrix and criterion weights are fuzzy Measured using linguistic values that are converted to values. Post-harvest in food value chains amount (PHL) diet researchers and policy to reduce these losses among classifiers It is increasingly discussed with the design of policies. First, reducing PHL is believed to improve food security because lower PHL will ensure more food availability at lower prices. Despite Comprehensive research and development of remedial techniques, Recommended control measures are limited There are Reducing post-harvest losses There is a lot of information published on related aspects fewer Effects of loss as fully as possible should be analyzed.

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