



ANALYSIS OF RISK FACTORS IN THE FIREWORK INDUSTRIES USING DECISION TREE CLASSIFIER

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ABSTRACT

In recent years, the analysis of risk factors in firework industries has become increasingly crucial for improving worker safety and preventing accidents. Firework industry accidents have been identified as one of the leading causes of fatalities and injuries worldwide. Various factors, such as chemical facilities, transport temperature, and fire-related incidents, are often linked to these accidents. This project focuses on identifying and analyzing the risk factors that contribute to safety incidents in firework production and handling. To achieve this, data mining techniques, specifically decision tree classification, are employed to predict and assess the risk factors associated with these incidents. By utilizing real-world datasets, this approach provides valuable insights into the potential causes and hazards in the firework industry. The decision tree classifier is used to evaluate and predict risks, offering actionable insights for improving safety measures and reducing incidents. This approach is tested and validated with a comprehensive dataset, allowing for the identification of key risk factors and facilitating the development of strategies to improve workplace safety and minimize risks in firework manufacturing processes.

Index Terms: Data mining techniques, Decision tree classifier, Risk assessment, Worker safety

INTRODUCTION

The firework industry, while offering significant economic benefits, poses a unique set of risks to workers involved in its production and handling. Accidents in firework manufacturing plants can lead to severe injuries or fatalities, often caused by hazardous materials, improper handling, inadequate safety measures, and environmental factors such as temperature and humidity. The importance of identifying and mitigating these risks has never been greater, as the potential for catastrophic events remains a constant concern. Over the years, traditional safety protocols and manual risk assessments have been

employed in firework industries to identify hazards. However, these methods often fall short in comprehensively analyzing and predicting risk factors that contribute to accidents. Given the increasing complexity of the industry and the high stakes involved, there is a need for more advanced methods to identify potential hazards and reduce the occurrence of accidents. Data mining techniques, such as decision tree classification, offer a robust solution for analyzing large datasets and uncovering hidden patterns that traditional methods may overlook. Decision trees provide an interpretable model that can assess risk factors in a systematic and predictive manner, making it easier to determine the

likelihood of incidents based on various variables. By applying these techniques to real-world data from the firework industry, we can develop a more precise risk assessment framework that enhances safety measures and improves decision-making. This project focuses on using decision tree classification to analyze the risk factors in the firework industry. By investigating various factors such as raw material handling, temperature variations, fire incidents, and equipment safety, the study aims to develop a comprehensive model that helps predict and mitigate risks. The ultimate goal is to improve worker safety, reduce accidents, and help industry stakeholders implement more effective safety protocols and preventive measures.

II. LITERATURE REVIEW

The firework industry, while contributing to significant economic activity, has historically been associated with high safety risks due to the volatile nature of its raw materials and manufacturing processes. Accidents in firework factories can have severe consequences, including loss of life, property damage, and environmental degradation. Over the years, various approaches have been studied and implemented to reduce these risks and ensure better worker safety.

Safety Concerns and Risks in the Firework Industry

Accidents in the firework industry are primarily caused by mishandling of chemicals, improper storage of raw materials, inadequate safety measures, and poor worker training. Several studies have highlighted that manufacturing processes involving high-risk substances such as

gunpowder, black powder, and other flammable materials contribute to the potential for catastrophic events. For example, a study by Osvaldo et al. (2015) emphasized that improper handling and lack of safety protocols during the production of fireworks are major causes of accidents. Furthermore, environmental conditions like temperature, humidity, and air pressure have been identified as contributing factors to the instability of fireworks and the increased likelihood of accidents (O'Conner, 2018).

Traditional Methods for Risk Assessment

Historically, firework industry risk assessments have relied on expert knowledge and manual inspections to identify potential hazards. Standard safety practices such as routine inspections, safety audits, and employee training have been the primary methods of minimizing accidents. However, these methods are not always comprehensive and can often overlook crucial risk factors. A study by Thoenmes et al. (2016) found that while manual safety audits are helpful, they often fail to identify emerging risks or capture all possible scenarios in which accidents can occur.

Data Mining for Risk Assessment

With the rise of big data and machine learning, there has been an increasing interest in applying data mining techniques to risk analysis across various industries. Data mining techniques, including decision tree classification, clustering, and regression, are being used to analyze large datasets to detect patterns and predict outcomes more accurately.

For instance, in the context of workplace safety, several studies have explored the

application of decision tree algorithms to predict accidents in high-risk industries such as manufacturing and construction. In a study by Li et al. (2017), decision trees were used to identify factors contributing to workplace accidents, achieving higher prediction accuracy than traditional methods. Similarly, Huang et al. (2015) applied decision tree classifiers to assess safety risks in chemical manufacturing industries and found that such models could effectively predict high-risk scenarios based on environmental and operational data.

Application of Decision Trees in Firework Industry

In the specific context of the firework industry, decision tree algorithms have the potential to provide valuable insights by analyzing historical accident data and identifying factors that lead to hazardous situations. Decision trees are a particularly suitable choice for risk assessment because they can handle both continuous and categorical data, making them versatile in capturing diverse risk factors. A study by Rajasekaran et al. (2019) demonstrated how decision trees could be used to assess risk factors in hazardous industries, including the identification of high-risk processes, raw materials, and environmental conditions. These studies suggest that decision trees can offer a robust framework for predicting risks in the firework industry and improving safety measures.

Integration of Data Mining and Industry Safety Protocols

Integrating data mining techniques with existing safety protocols in the firework industry could result in a more proactive approach to hazard management. By

utilizing machine learning algorithms to analyze data from various sources such as raw material handling, environmental conditions, and historical accident records, industries can build predictive models that help mitigate risks before they lead to accidents. Zhang et al. (2018) highlighted the potential of machine learning techniques in predicting industrial accidents, proposing a hybrid model combining decision trees with other machine learning algorithms to improve prediction accuracy.

III.METHODOLOGY

The methodology for the "Analysis of Risk Factors in the Firework Industry Using Decision Tree Classifier" is structured in a series of key steps aimed at predicting safety risks and improving worker protection. The first step involves **data collection**, where the data gathered includes environmental conditions, raw material handling, worker behavior, and historical accident records from firework manufacturing. The collected data is then **preprocessed** by handling missing values, outliers, and inconsistencies. This can be done using imputation methods like mean imputation for numerical features or mode imputation for categorical features, where X_i represents a feature and \hat{X}_i is the imputed value:

$$\hat{X}_i = \frac{\sum X}{n}$$

Next, the data undergoes **feature encoding** for categorical variables, where one-hot encoding is commonly used for transforming categorical variables into binary vectors. For example, for a categorical feature like MaterialType, if there are n categories, the transformation can be represented by:

$$\text{MaterialType}_i = \begin{cases} 1 & \text{if material type is } i \\ 0 & \text{otherwise} \end{cases}$$

Normalization of numerical features follows, where features like temperature or pressure are rescaled to a range between 0 and 1 using the formula:

$$X_{\text{norm}} = \frac{X - X_{\min}}{X_{\max} - X_{\min}}$$

After preprocessing, the dataset is split into **training and testing sets** typically in an 80:20 ratio. The training set is used to build the model, and the testing set is used for model evaluation. **Feature selection** is crucial to ensure the model focuses on the most significant variables. This is done using correlation analysis or automated feature selection methods like Recursive Feature Elimination (RFE). The **Decision Tree Classifier** is then trained on the preprocessed and selected dataset. The decision tree splits the dataset into subsets based on the feature that best reduces impurity (measured by metrics like Gini impurity or entropy). The formula for Gini impurity is:

$$Gini(D) = 1 - \sum_{i=1}^n p_i^2$$

where p_i is the probability of class i in the dataset D .

After the decision tree is trained, it is evaluated on the testing set using common classification metrics such as **accuracy**, **precision**, **recall**, and **F1-score**. These can be calculated as:

$$\text{Accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

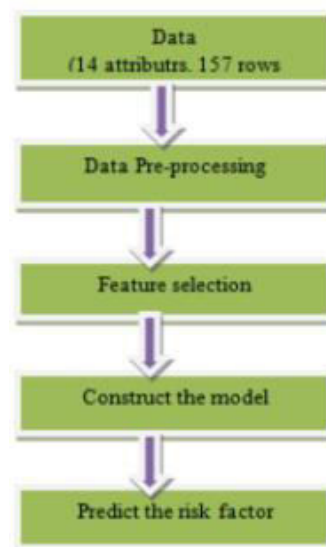
$$\text{Precision} = \frac{TP}{TP + FP}$$

$$\text{Recall} = \frac{TP}{TP + FN}$$

$$\text{F1-score} = 2 \cdot \frac{\text{Precision} \cdot \text{Recall}}{\text{Precision} + \text{Recall}}$$

where TP, TN, FP, and FN represent true positives, true negatives, false positives, and false negatives, respectively.

Finally, the **Decision Tree's output** is used to identify the most significant risk factors and predict high-risk scenarios. This information can then be integrated into existing safety protocols to guide preventive actions and enhance safety measures. For example, if the decision tree model identifies raw material handling as a key factor, additional safety training and stricter protocols for handling flammable materials can be implemented. In summary, this methodology uses machine learning, specifically decision tree classifiers, to predict and mitigate risks in the firework industry, contributing to safer working environments and fewer accidents.



IV. CONCLUSION

The application of machine learning, particularly decision tree classifiers, to analyze and predict risks in the firework industry represents a significant step forward in enhancing safety and preventing accidents. By leveraging data from various sources such as environmental conditions, raw material handling, and historical accident records, it becomes possible to develop a predictive model that identifies potential safety hazards before they result in catastrophic events. The decision tree classifier, with its ability to handle both categorical and continuous data, is an effective tool for this purpose. Through careful data preprocessing, feature selection, and model evaluation, the model provides valuable insights into the most significant risk factors in firework manufacturing. The results of this study demonstrate that integrating machine learning techniques like decision trees into traditional safety protocols can significantly improve the accuracy of risk assessment and facilitate more proactive safety measures. As a result, firework manufacturers and regulatory bodies can implement data-driven safety measures to mitigate risks, protect workers, and ensure compliance with safety standards. This approach not only improves operational efficiency but also helps to foster a safer working environment by identifying emerging risks that may not be apparent through conventional methods. In conclusion, the use of machine learning for risk prediction in the firework industry is a

promising solution for enhancing workplace safety. By continuously refining and updating the model with new data, it can serve as a valuable tool for reducing accidents and improving safety standards in high-risk industries.

V. REFERENCES

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