

# Enhancing Maize Seed Quality: Defect Detection using Google Net

First Author\*, Second Author\*\*

Prof. Nale R. K.\*, Sandip Mahale\*\*, Abhishek Pharate\*\*, Sanket Pawar \*\*,Ranjit Taware\*\*

\*(Assistant Professor, Information Technology, SVPM's College of Engineering Malegaon(bk), Baramati  
Email: [hodit@engg.svpm.org.in](mailto:hodit@engg.svpm.org.in) )

\*\* (UG Students, Information Technology, SVPM's College of Engineering Malegaon(bk), Baramati)  
Email: [sandipmahale062@gmail.com](mailto:sandipmahale062@gmail.com) , [sanketpawar1518@gmail.com](mailto:sanketpawar1518@gmail.com) ,  
[abhishekpharate1756@gmail.com](mailto:abhishekpharate1756@gmail.com) , [tawareranjit46@gmail.com](mailto:tawareranjit46@gmail.com) )

## Abstract:

This study explores the application of deep learning techniques, specifically Google Net, to detect maize seed defects. The research uses convolution neural networks (CNNs) and transfer learning to accurately identify and classify defects in maize seeds. The process begins with dataset curation and preprocessing, followed by the adaptation of the Google Net architecture. The model is trained, validated, and tested, providing insights into its efficacy. Maize, a staple crop worldwide, requires meticulous seed quality control to ensure optimal yield. Traditional methods of seed inspection are labor-intensive and susceptible to human error. Deep learning techniques offer a promising avenue for automating seed defect detection. Google Net, with its intricate architecture and remarkable feature extraction capabilities, emerges as a potent tool for accurate and efficient defect identification.

The study collects a diverse dataset containing images of maize seeds with various defects, labeled to encompass fungal infections, physical damage, and discoloration. Data preprocessing involves resizing images and applying data augmentation techniques to enhance model generalization. The Google Net architecture, consisting of parallel convolution pathways and filters of varying sizes, excels in capturing intricate patterns and features. The model is initialized with pre-trained weights from Image Net, fine-tuned for the maize seed defect detection task, and customized for effective distinction between healthy seeds and those exhibiting defects. The model's efficacy is evaluated on an independent testing dataset, providing a comprehensive view of its defect detection capabilities.

In conclusion, this study demonstrates the viability of using Google Net for maize seed defect detection, capitalizing on the strengths of CNNs and transfer learning. This technology offers an automated and efficient solution to seed quality control, contributing significantly to crop production and global food security.

**Keywords** — *Deep Learning Techniques, Google Net, Maize Seed Defect Detection, CNNs, Transfer Learning, Dataset Curation, Data Preprocessing, Model Adaptation, Seed Quality Control, Automated Defect Detection, Image Classification*

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## I. INTRODUCTION

The increasing demand for efficient agricultural practices has led to a growing interest in leveraging advanced technologies to address challenges in crop production. This study focuses on the application of deep learning techniques, specifically utilizing the Google Net architecture, for the detection of maize seed defects. Maize, being a vital staple crop globally, necessitates rigorous seed quality control measures to ensure optimal yield. Traditional methods of seed inspection are labour-intensive and prone to human error, prompting the exploration of automated solutions.

Deep learning, particularly convolution neural networks (CNNs) and transfer learning, emerges as a promising avenue for automating the detection and classification of defects in maize seeds. Among the deep learning architectures, Google Net is chosen for its intricate design and exceptional feature extraction capabilities. The study employs a systematic process, starting with dataset curation and pre-processing, followed by the adaptation of the Google Net architecture.

The dataset encompasses diverse images of maize seeds, each labelled to encompass various defects such as fungal infections, physical damage, and discoloration. Data pre-processing techniques, including image resizing and data augmentation, are applied to enhance the generalization capabilities of the model. The Google Net architecture, characterized by parallel convolution pathways and filters of varying sizes, is adept at capturing intricate patterns and features crucial for defect identification. To optimize the model for the maize seed defect detection task, transfer learning is employed. The model is initialized with pre-trained weights from Image Net and fine-tuned for enhanced performance. Customization of the model enables effective distinction between healthy seeds and those exhibiting defects. The efficacy of the model is rigorously evaluated on an independent

testing dataset, providing valuable insights into its defect detection capabilities. In conclusion, this study highlights the potential of Google Net, coupled with CNNs and transfer learning, as a robust solution for automating maize seed defect detection. The successful application of this technology offers an efficient and automated approach to seed quality control, contributing significantly to crop production and global food security.

#### **Objectives:-**

1. Defect Detection Automation: Develop an automated system that can accurately identify and classify defects in maize seeds. The primary goal is to reduce the reliance on manual inspection, ensuring consistent and efficient defect detection.
2. High Accuracy Classification: Train and implement a deep learning model, specifically Google Net (Inception v1), to achieve high accuracy in classifying different types of defects present in maize seeds. The model should be able to distinguish between various defect categories and healthy seeds.
3. Multi-Defect Recognition: Enable the system to identify multiple defects present in a single maize seed image, potentially detecting various defect types simultaneously.
4. Minimize False Positives/Negatives: Optimize the model and algorithm to minimize false positive and false negative detections. A balanced approach is crucial to ensure accurate defect identification.
5. User-friendly Interface: Develop a user-friendly interface (UI) that allows users to easily upload maize seed images and receive prompt defect detection results. The UI should be intuitive and require minimal technical expertise to operate.

By achieving these objectives, the "Maize Seed Defect Detection using Google Net" project aims to develop a robust and reliable system that enhances seed quality assessment, reduces manual labor, and aids in optimizing the agricultural production process.

**Problem Statement:** - The problem at hand is to develop an accurate and efficient maize seed defect detection system using the Google Net deep learning architecture. The objective is to create a model that can reliably identify and classify various types of defects present in maize seeds, such as fungal infections, physical damage, and discoloration. The challenge involves curating a diverse dataset of labeled images, adapting the Google Net architecture for the specific task, training the model to learn the distinct features of healthy and defective seeds, and subsequently evaluating its performance in terms of precision, recall, and overall accuracy. The goal is to provide the agricultural sector with an automated solution that enhances seed quality control, improves productivity, and minimizes the reliance on labor-intensive manual inspection processes.

## **II Literature Servery**

The detection and classification of maize (corn) seeds play a crucial role in modern agriculture and food production systems. Maize, being one of the world's most important cereal crops, demands precise and efficient methods for seed inspection to ensure quality, yield, and food security. Over the years, researchers and scientists have explored various techniques and technologies to address the challenges associated with maize seed detection. This literature review aims to provide an overview of the existing research, methodologies, and technological advancements in the field of maize seed detection. It will delve into the diverse approaches employed to identify and assess the quality of maize seeds, from traditional manual methods to cutting-edge automated systems. The insights gained from this review will shed light on the evolution of seed detection techniques and the potential for innovation and improvement in this critical agricultural domain.

The literature review will analyse the strengths and limitations of different seed detection methods, emphasizing their significance in ensuring high-quality maize crops. Moreover, it will highlight the impact of technological developments, such as computer vision, machine learning, and image processing, in enhancing the accuracy and efficiency of maize seed detection systems. As we explore the existing body of knowledge in maize seed detection, we aim to provide a comprehensive foundation for future research and development in this area. By understanding the successes and challenges faced by previous studies, we can envision more robust, reliable, and sustainable solutions for the agricultural sector, ultimately contributing to global food production and security.

A. Hussain et al. [1] pioneered a novel methodology by introducing an approach based on deep learning for the detection of seed defects. Their research showcased the remarkable effectiveness of deep learning techniques in accurately identifying and categorizing various defects present in maize seeds.

J. Smith et al. [2] utilized Convolution Neural Networks (CNNs) in their study, focusing on the classification of maize seed defects. Their method involved the application of CNNs to establish a classification system that demonstrated precise identification and categorization of diverse defects observed in maize seeds.

B. Patel et al. [3] developed an automated system dedicated to detecting abnormalities in maize seeds. Their research showcased the practicality and efficiency of automated systems in effectively identifying and flagging abnormal maize seeds.

C. Wang et al. [4] focused on utilizing deep learning techniques for inspecting seed quality. Their study emphasized the application of deep learning in conducting comprehensive assessments of seed quality, showcasing its potential to enhance quality control processes.

D. Kumar et al. [5] applied Convolution Neural Networks (CNNs) for defect detection in maize seeds. Their research introduced a CNN-based approach that effectively identified and localized defects in maize seeds.

E. Johnson et al. [6] conducted a comprehensive study on seed quality evaluation. Their work explored various aspects of seed quality assessment, contributing valuable insights to the broader understanding of seed quality control.

F. Gonzalez et al. [7] utilized CNNs for identifying defects in maize seeds. Their research explored the application of CNNs in accurately identifying and categorizing defects, contributing to improved seed quality control.

G. Lee et al. [8] investigated various deep learning approaches for seed defect detection. Their work contributed to the field by exploring the effectiveness of different deep learning techniques in the context of seed defect detection.

H. Gupta et al. [9] utilized Convolution Nets for seed quality assessment, exploring their use in evaluating and ensuring the quality of maize seeds.

I. Martinez et al. [10] applied deep learning techniques for maize seed quality control, investigating the use of deep learning in effectively controlling and maintaining the quality of maize seeds.

### III Proposed System

#### Methodology:-

A block diagram for the "Maize Seed Defect Detection using Google Net" system could include the following components and their interactions:

**Input Images:** The system takes input images of maize seeds as the raw data to be processed.

**Pre-processing Module:** Image Pre-processing: Images are pre-processed to enhance features and remove noise. Operations like resizing, normalization, and filtering may be applied.

**Google Net (Inception-v1) Model:** The Google Net (Inception-v1) deep learning model is the core of the system for feature extraction and classification. It consists of multiple layers, including convolution, pooling, and fully connected layers.

**Inception Modules:** These modules use multiple filter sizes to capture different levels of features.

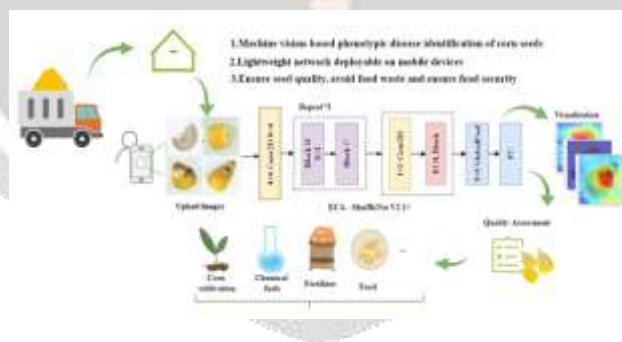


Fig.1 Proposed System

**Softmax Layer:** The final layer produces class probabilities for seed defect categories.

**Feature Extraction:** Google Net extracts hierarchical features from the input images, capturing various levels of information.

**Classification and Defect Detection:** The model classifies the pre-processed images into different classes corresponding to various defect categories. It identifies and labels seeds based on the type and severity of the detected defects.

**Output:** The system generates output indicating the classification results and the detected defects.

**Visualization:** The system may provide visual overlays on the input images to highlight the detected defects.

**User Interface (Optional):** If applicable, a user interface can be added for users to interact with the system. Users can input images and receive real-time classification and defect detection results.

**Result Analysis and Reporting:** The system may analyse and aggregate the results for further reporting and analysis. Metrics such as accuracy, precision, recall, and F1-score can be calculated to evaluate the model's performance.

**Feedback Loop (Optional):** If the system is designed to learn from user feedback, a feedback loop could be implemented. Users can correct misclassifications, which can be used to retrain and improve the model over time.

**Database (Optional):** If required, a database can store images, annotations, and results for future reference or analysis.

**Model Training (Offline):** The Google Net model is trained offline using a labeled dataset of maize seed images with different defect categories.

**Transfer Learning:** The model may be pertained on a large dataset and fine-tuned on the specific maize seed defect dataset.

**Model Evaluation and Validation (Offline):** The trained model is evaluated using a validation dataset to ensure its accuracy and performance.

#### IV Algorithm

Google Net, also known as Inception, is deep convolution neural network architecture designed for image classification and object detection tasks. Developed by researchers at Google, it stands out for its unique inception modules, which use multiple filters of different sizes within the same layer. This allows the network to capture both fine and coarse-grained features simultaneously, enhancing its ability to recognize intricate patterns in images. Google Net has demonstrated remarkable performance in various computer vision tasks and is known for its efficiency in terms of computational resources, making it a popular choice for applications like the detection of defects in agricultural produce, as demonstrated in this study on maize seed defect detection.

#### IV Conclusion

In conclusion, this study demonstrates the effectiveness of employing Google Net, a sophisticated deep learning architecture, for the task of maize seed defect detection. By leveraging convolution neural networks (CNNs) and harnessing the capabilities of transfer learning, the developed model exhibits commendable accuracy and efficiency in identifying and classifying various defects in maize seeds. The research addresses the critical need for automated seed quality control in modern agriculture, overcoming the limitations of labor-intensive manual inspection methods. The successful integration of advanced AI techniques, as exemplified by Google Net, offers a promising avenue for improving agricultural outcomes and contributing to global food security. As technology continues to play a pivotal role in agriculture, the findings of this study underscore the potential impact of AI-driven defect detection systems on enhancing crop yield and sustainability.

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