

Wood Quality Analyzing System

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Abstract - This paper presents a comprehensive approach to optimizing furniture design processes and facilitating wood identification through the development of a mobile application. The application is designed to provide users with intuitive tools and resources for informed decision-making in furniture design and wood selection. The first component of the mobile application focuses on optimizing furniture design through furniture size and quality consideration. This encompasses the creation of an intuitive user interface that enables users to specify their furniture design preferences, including material selection criteria. A comprehensive wood database is meticulously curated, furnishing users with comprehensive details on wood sizes, qualities, and thicknesses, thereby aiding in material selection and waste reduction initiatives. Algorithms are employed to recommend optimal wood sizes, considering factors such as material selection, waste reduction, and cost analysis. Visual representation tools are included, allowing users to preview furniture designs with different wood options, facilitating informed decision-making. Furthermore, feedback mechanisms have been integrated to enable users to provide feedback on recommended wood sizes and qualities, thereby facilitating continuous enhancement and refinement of recommendations. The second component aims to identify the type of wood by its appearance. Image processing features have been developed to precisely extract visual characteristics from wood sample images, thereby aiding in the identification of wood species. Machine learning models are trained to classify wood species based on visual characteristics obtained from images, thereby enhancing the precision of identification. User interaction tools have been designed to ensure ease of use and accessibility for users in capturing and submitting wood sample images. Within the application, educational resources and materials are provided to help users learn about different wood species, aiding their understanding and selection process.

Keywords: Furniture Design Optimization, Wood Identification, Mobile Application Development, Material Selection, User Feedback Integration.

I. INTRODUCTION

In the realm of furniture design, it is imperative to be able to make informed decisions regarding material selection in order to achieve optimal outcomes. This paper introduces a groundbreaking mobile application designed to revolutionize both the furniture design process and wood identification and selection within the industry. The mobile application is designed to revolutionize both the furniture design process and wood identification and selection within the industry. By integrating advanced features into this application, this application aims to streamline design decision-making and enhance user experience.

This innovative application focuses on optimizing furniture design through meticulous consideration of wood size and quality. It boasts an intuitive user interface that enables users to specify their design preferences, including material selection criteria, with ease and precision. The interface is complemented by a comprehensive wood database that furnishes users with comprehensive details on wood sizes, qualities, and thicknesses. By allowing access to this wealth of information, the application facilitates informed decision-making during the material selection process, thereby contributing to waste reduction efforts and promoting sustainable practices. The core of the application's functionality lies in a sophisticated set of algorithms designed to recommend optimal wood sizes based on a holistic analysis of material selection criteria, waste reduction goals, and cost analysis considerations. These algorithms utilize user preferences and project prerequisites to generate personalized recommendations that are tailored to each user's distinct requirements.

Furthermore, the application incorporates powerful visual representation tools that allow users to preview furniture designs with different wood options. The application provides users with a visual representation of their design choices, enabling them to make more informed decisions about material selection and design aesthetics. In addition to its robust recommendation capabilities, the application boasts integrated feedback mechanisms that enable users to provide

feedback on recommended wood sizes and qualities. This feedback loop serves as a crucial mechanism for continual enhancement and refinement, ensuring that the application evolves to better cater to the requirements and preferences of its users over time. Due to emerging trends and technological advances, the second component of the mobile application is dedicated to wood identification based on visual appearance. The application leverages advanced image processing features and machine learning models to precisely extract visual characteristics from wood sample images, facilitating accurate wood species identification. User-friendly interaction tools make it easier for users with varying levels of technical expertise to capture and submit wood sample images.

Furthermore, the application provides educational resources that enable users to enhance their comprehension of diverse wood species, thereby facilitating more informed decisions in furniture design and material selection. The mobile application sets a new standard for excellence both in furniture design and wood identification and selection. The application leverages advanced technologies and user-centric design principles to enable users to make informed decisions and achieve optimal outcomes in their design projects.

II. BACKGROUND & LITERATURE REVIEW

Further investigation into optimizing furniture design processes reveals the multifaceted nature of achieving efficiency [1]. Material selection involves considerations of durability, aesthetics, and environmental impact, which drive the need for sustainable practices [2]. It is essential to incorporate sustainability standards into design processes in order to minimize ecological footprints and meet consumer demand for eco-friendly products.

In the field of wood identification, advances in image processing and machine learning offer promising avenues for improving accuracy [3]. These technologies streamline the identification procedure, enabling a more informed material selection based on factors such as wood species and quality. Furthermore, they contribute to sustainability efforts by enabling the use of responsibly sourced materials.

The examination of user experience in mobile applications demonstrates the crucial significance of consumer insights in influencing design decisions [4]. The insights gleaned from studies on online furniture consumption inform user-centric design strategies, ensuring intuitive interfaces and seamless interactions. Accessibility considerations further enhance usability by catering to diverse user requirements and preferences.

Process optimization techniques, such as process mapping, play a crucial role in streamlining manufacturing

operations [5]. These methods improve productivity and resource utilization in furniture manufacturing processes by identifying bottlenecks and inefficiencies. This optimization contributes to cost reduction and improves overall operational efficiency.

To deliver personalized solutions, it is crucial to understand user preferences. Research into demand factors sheds light on changing consumer preferences, guiding the development of tailored design recommendations [6]. Innovative user-centric design approaches drive innovation and differentiation in the competitive landscape of furniture design.

III. RESEARCH GAP

One significant obstacle lies in the disparate nature of the two components. While the initial component focuses on optimizing furniture design through considerations such as size, quality, and material selection, the second component revolves around identifying wood species based on visual characteristics. At present, there exists a deficiency in synergy among these components, with minimal interaction or integration between them. Moreover, the present framework predominantly treats wood identification as a distinct feature within the application, rather than as a fundamental component of the furniture design optimization process. The utilization of insights obtained through wood identification to enhance the overall design recommendations provided to users is limited by this approach. To address this deficiency, future research could explore strategies for seamlessly incorporating wood identification insights into the furniture design optimization process. This may entail the development of algorithms or models that incorporate data from wood identification to refine and personalize the design recommendations provided to users. For example, information about the specific species identified in a user's sample image could be used to tailor recommendations for compatible wood types or design styles.

Furthermore, it would be beneficial to provide more transparent and intuitive connections between the wood identification and furniture design components of the application. This could encompass visual representations or elucidations of the impact of wood identification data on the design recommendations provided to users, enabling them to make more informed decisions regarding their furniture design choices. To address the research gap identified, it would require a more holistic approach to the design and functionality of the mobile application, one that integrates wood identification seamlessly into the furniture design optimization process. This would enhance the user's experience and decision-making capabilities.

IV. RESEARCH PROBLEM

The disjointed nature of the wood identification and furniture design optimization components leads to inefficiency in decision-making. Users are compelled to navigate between different features or interfaces within the application, resulting in a fragmented user experience. This fragmentation can impede users ability to make timely and informed decisions about their furniture designs, potentially causing frustration and dissatisfaction.

Furthermore, the lack of seamless integration limits the application's ability to offer personalized recommendations tailored to individual users preferences and project requirements. Instead, users may encounter generic design suggestions that fail to consider the distinctive characteristics of the identified wood species. Consequently, the application may encounter difficulties in accommodating the diverse requirements and preferences of its user base, thereby diminishing its overall efficacy.

Additionally, the absence of integration between wood identification and design optimization means that valuable insights derived from wood analysis are not fully utilized. It is possible that crucial information regarding the quality or attributes of a particular wood species may not be adequately incorporated into the algorithm responsible for suggesting optimal wood sizes or design styles. This oversight represents a missed opportunity to provide tailored and efficient design solutions, compromising the application's utility and effectiveness.

Ultimately, the fragmented user experience resulting from the absence of integration may result in user dissatisfaction and disengagement. Users encountering obstacles or inconsistencies may become frustrated with the application's usability, leading them to abandon the platform altogether. The impact of user dissatisfaction on the application's success may include a reduction in user satisfaction and retention rates. The research issue necessitates a comprehensive examination of the impact of the absence of integration between wood identification and furniture design optimization on user experience, personalization efforts, utilization of insights, and overall user engagement with the application. These insights are critical for devising effective strategies to bridge identified gaps and enhance the application's functionality and user satisfaction.

V. OBJECTIVES

A) For Optimizing Furniture Design through Wood Size and Quality Consideration:

1) *Develop an Intuitive User Interface:*

Create a user-friendly interface allowing users to specify furniture design preferences, including material selection criteria.

2) *Create a Comprehensive Wood Database:*

Compile a detailed database with information on wood sizes, qualities, and thicknesses, aiding in material selection and waste reduction efforts.

3) *Implement Algorithms for Wood Optimization:*

Develop algorithms to recommend optimal wood sizes considering factors like material selection, waste reduction, and cost analysis.

4) *Incorporate Visual Representation Tools:*

Integrate visual tools enabling users to preview furniture designs with different wood options, facilitating informed decision-making.

5) *Integrate Feedback Mechanisms:*

Incorporate mechanisms for users to provide feedback on recommended wood sizes and qualities, allowing for continuous improvement and refinement of recommendations.

B) For Identifying the Type of Wood by Its Appearance:

1) *Implement Image Processing Features:*

Develop image processing features to extract visual features from wood sample images accurately, aiding in wood species identification.

2) *Train Machine Learning Models:*

Train machine learning models to classify wood species based on visual characteristics extracted from images, enhancing the accuracy of identification.

```
[ ] # Build the model
model = Sequential()

model.add(Conv2D(32, (3, 3), activation='relu', input_shape=(img_size, img_size, 3)))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Conv2D(128, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Flatten())
model.add(Dense(256, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(len(labels), activation='softmax'))

# Compile the model
model.compile(optimizer=Adam(), loss='categorical_crossentropy', metrics=['accuracy'])

# Display the model summary
model.summary()

# Train the model
history = model.fit(X_train, y_train, epochs=10, validation_data=(X_val, y_val))
```

Figure 1: Build the model

3) Design User Interaction Tools:

Create user-friendly tools for capturing and submitting wood sample images, ensuring ease of use and accessibility for users.

4) Provide Educational Resources:

Offer educational resources and materials to help users learn about different wood species, aiding in their understanding and selection process.

VI. METHODOLOGY

A) Optimizing Furniture Design through Wood Size and Quality Consideration

1) User Interface Development:

Approach: Employ Android Studio to design and implement a user-friendly mobile application interface.

Implementation: Create interactive screens enabling users to input furniture design preferences, including material selection criteria.

Tools: Android Studio for interface design, Java/Kotlin for backend logic.

2) Wood Database Compilation:

Approach: Compile a comprehensive database containing detailed information on various wood types, sizes, qualities, and thicknesses.

Implementation: Utilize a CSV file to store wood-related data and load it into a panda Data Frame for efficient data handling.

```

#mount the google drive
from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

[ ] import matplotlib.pyplot as plt
import seaborn as sns
import tensorflow as tf
from tensorflow.keras.models import Sequential, Model
from tensorflow.keras.layers import Dense, Conv2D, MaxPool2D, Flatten, Dropout, MaxPooling2D
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report, confusion_matrix
from tensorflow.keras.applications import VGG19, ResNet50
import cv2
import os
import random
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
    
```

Figure 2: Generate CSV File

Tools: Python with pandas' library for data manipulation and storage.

3) Algorithm Development for Wood Optimization:

Approach: Develop algorithms to recommend optimal wood sizes considering factors like material selection, waste reduction, and cost analysis.

Implementation: Utilize the DecisionTreeClassifier algorithm to predict optimal wood sizes based on input criteria.

Tools: Python for algorithm development, scikit-learn library for machine learning functionalities.

4) Visual Representation Tool Integration:

Approach: Integrate visual tools into the mobile application to enable users to preview furniture designs with different wood options.

Implementation: Develop features allowing users to visualize furniture designs using selected wood types and configurations.

Tools: Android Studio for interface design and implementation of visualization functionalities.

5) Feedback Mechanism Incorporation:

Approach: Implement mechanisms within the application for users to provide feedback on recommended wood sizes and qualities.

Implementation: Develop interactive elements to collect user feedback on recommended wood options and incorporate it for continuous improvement.

Tools: Android Studio for interface design and implementation of feedback mechanisms.

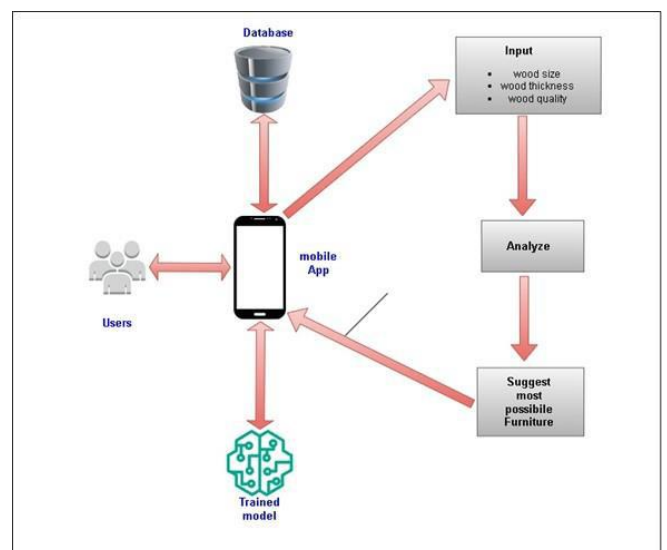


Figure 3: Component Diagram

B) For Identifying the Type of Wood by Its Appearance

Develop image processing features for accurately extracting visual features from wood sample images to identify wood species. For this step, the following activities are needed:

Approach: The image processing is performed using Python libraries, including OpenCV and scikit-image.

Implementation:

Pre-process Images: Preprocess wood sample images is performed in order to improve the quality of images; normalization and noise filtration are applied.

Feature Extraction: Extract the most relevant visual features from wood sample images; include texture, color, and grain patterns.

Image Segmentation: Perform image segmentation of wood sample images to identify regions of interest, which are the most interesting for analyzing.

Feature representation: Represent extracted features of visual components in suitable format for further analysis and model training. Tools: Python, OpenCV, scikit-image.

1) Training Machine Learning Models

Develop and train machine learning models that can classify wood species based on visual features extracted from the images. The following steps should be performed:

Approach and type of learning: Use supervised learning to develop classification models based on feature extraction from images

Implementation process:

Data collection: Collect the wood color sample image dataset and label them according to their species.

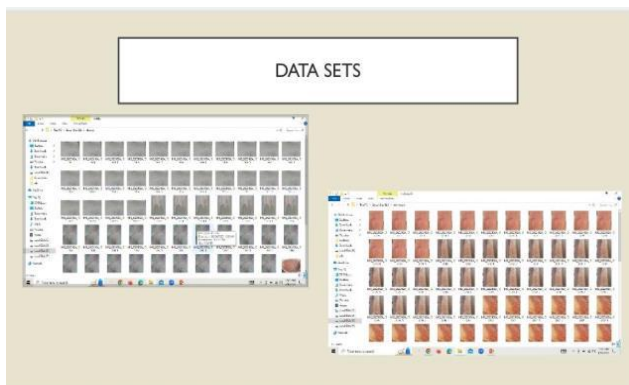


Figure 4: Data Set

Data preprocessing: Convert the dataset to training and testing sets and perform preprocessing steps.

Model selection: Select appropriate machine learning classifications such as CNN-based or SVM-based scores.

Model training: train the model with the OTTO wood alignment image set and test the model on the test sets.

Model optimization: Alter some hyperparameters and model architecture to increase accuracy. Tools: Python, tensorflow, scikit-learn.

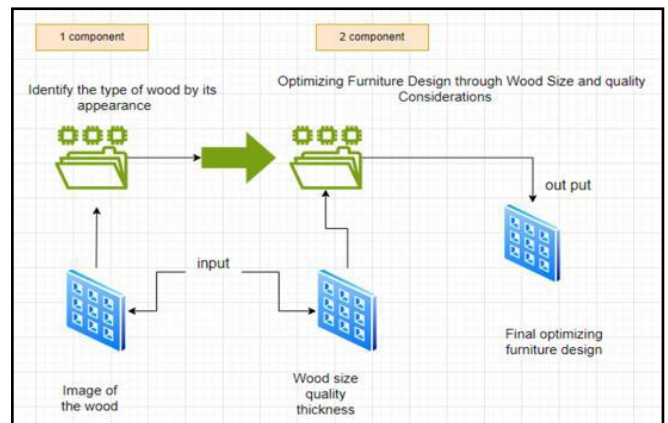


Figure 5: Component Diagram

2) Designing User Interaction Tools:

Create user-friendly tools for capturing and submitting wood sample images, ensuring ease of use and accessibility for users.

Approach: Design intuitive interfaces and functionalities to facilitate image capture and submission within the mobile application.

Implementation:

User Interface Design: Develop a visually appealing and intuitive interface for the mobile application, allowing users to easily capture and submit wood sample images.

Image Capture Functionality: Implement features enabling users to capture high-quality images of wood samples using the device's camera.

Submission Mechanism: Provide options for users to submit captured images for wood species identification and analysis within the application.



Figure 6: Prediction

Tools: Android Studio, Java/Kotlin.

Below is the accuracy of the component.

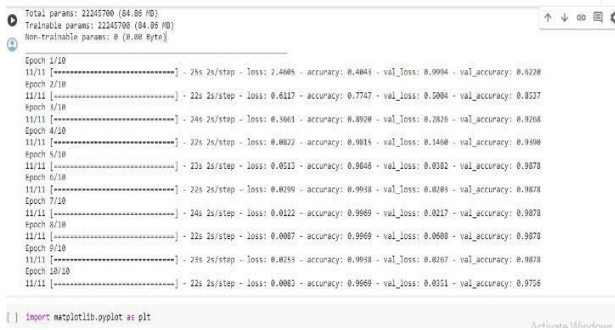


Figure 7: Check the accuracy

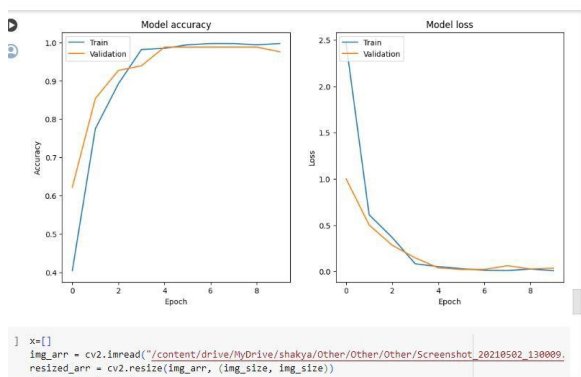


Figure 8: Accuracy Chart

VII. RESULTS AND DISCUSSION

After the optimization of furniture design through consideration of wood size and quality involves several key components. Firstly, the development of a user-friendly interface with Android Studio allows users to easily specify their furniture design preferences, including criteria for material selection. This interactive platform facilitates user engagement and informed decision-making. Furthermore, the compilation of a comprehensive wood database containing detailed information on various wood types, sizes, qualities, and thicknesses enables efficient material selection and waste reduction efforts. By utilizing tools such as CSV files and

Pandas Data Frames, the database guarantees seamless data handling and retrieval.

Furthermore, algorithms are developed to recommend optimal wood sizes, taking into consideration factors such as material selection, waste reduction, and cost analysis. The utilization of the Decision Tree Classifier algorithm facilitates precise recommendations that are tailored to user preferences. Furthermore, the mobile application integrates visual tools that allow users to preview furniture designs with different wood options. Using this visual representation enhances user engagement and aids in decision-making processes. Furthermore, feedback mechanisms are incorporated to allow users to provide insight into recommended wood sizes and qualities, fostering continuous improvement and refinement.

On the contrary, in the identification of wood types based on their appearance, image processing features have been developed to precisely extract visual characteristics from wood sample images. Techniques such as feature extraction and preprocessing improve the accuracy of wood species identification. On labeled datasets, machine learning models are then trained using supervised learning techniques to classify wood species based on extracted visual features. Ultimately, user interaction tools have been developed to facilitate the capture and submission of wood sample images within the mobile application, guaranteeing accessibility and ease of use for users contributing to the identification of wood species.

VIII. CONCLUSION

In conclusion, it is crucial to optimize furniture design through considerations of wood size and quality, as well as the identification of wood types by appearance, in the woodworking and furniture industries. Modern technologies and methodologies, such as user-friendly interfaces, comprehensive wood databases, algorithmic recommendations, visual representation tools, image processing, and machine learning, can be used to make significant advances in these areas.

The implementation of intuitive user interfaces and comprehensive wood databases streamlines the furniture design process, enabling users to make informed decisions while minimizing waste and maximizing efficiency. Additionally, the integration of algorithms for wood optimization and visual representation tools enhances the decision-making process by providing users with accurate recommendations and realistic previews of furniture designs.

Similarly, the implementation of image processing features and machine learning models for wood species identification greatly improves the accuracy and efficiency of

this task. By allowing users to capture and submit wood sample images through user-friendly interfaces, the process will be more accessible and streamlined.

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REFERENCES

- [1] Optimization of furniture combination design and space configuration based on graph theory. (2024,

February 23). ResearchGate. [Online]. Available: https://www.researchgate.net/publication/375330832_Optimization_of_furniture_combination_design_and_space_configuration_based_on_graph_theory

- [2] Muhammad Suandi, M. E. (2022). A Review on Sustainability Characteristics Development. MDPI. [Online]. Available: <https://www.mdpi.com/2071-1050/14/14/8748>
- [3] Development of mobile-based application for practical wood identification. (n.d.). ResearchGate. [Online]. Available: https://www.researchgate.net/publication/346154842_Development_of_mobile-based_application_for_practical_wood_identification
- [4] Artificial intelligence and smart vision for building. (n.d.). ScienceDirect. [Online]. Available: <https://www.sciencedirect.com/science/article/pii/S0926580522003132>
- [5] Zhang, J. (2022). Optimization of Network Furniture Management System. Hindawi. [Online]. Available: <https://www.hindawi.com/journals/mpe/2022/9698853/>
- [6] Lyu, C., & Chen, L. (n.d.). Applying a Hybrid Kano/Quality Function Deployment Integration. Semantic Scholar. [Online]. Available: <https://www.semanticscholar.org/paper/Applying-a-Hybrid-Kano-Quality-Function-Deployment-Lyu-Chen/19eddb545efaa8d618169d6a9abc49dcb27cae90>

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