

INTERNSHIP PROGRAM

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Presentation TOPICS





1st Topic

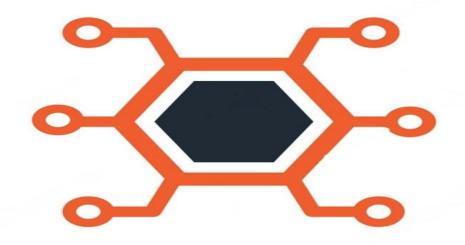
LLM and its application

Introduction to Open Source LLMs

Open source LLMs (large language models) are a type of artificial intelligence (AI) that can be used for a variety of tasks in data science. They are trained on massive datasets of text and code, and can be used to generate text, translate languages, write different kinds of creative content, and answer your questions in an informative way

2nd Topic

UNSUPERVISED LEARNING



Unsupervised Learning

- A cluster is represented by a single point, known as centroid (or cluster center) of the cluster.

Types of clustering:

- K-means, Fuzzy C, hierarchical

Similarity functions:

- Euclidean distance, Manhattan distance

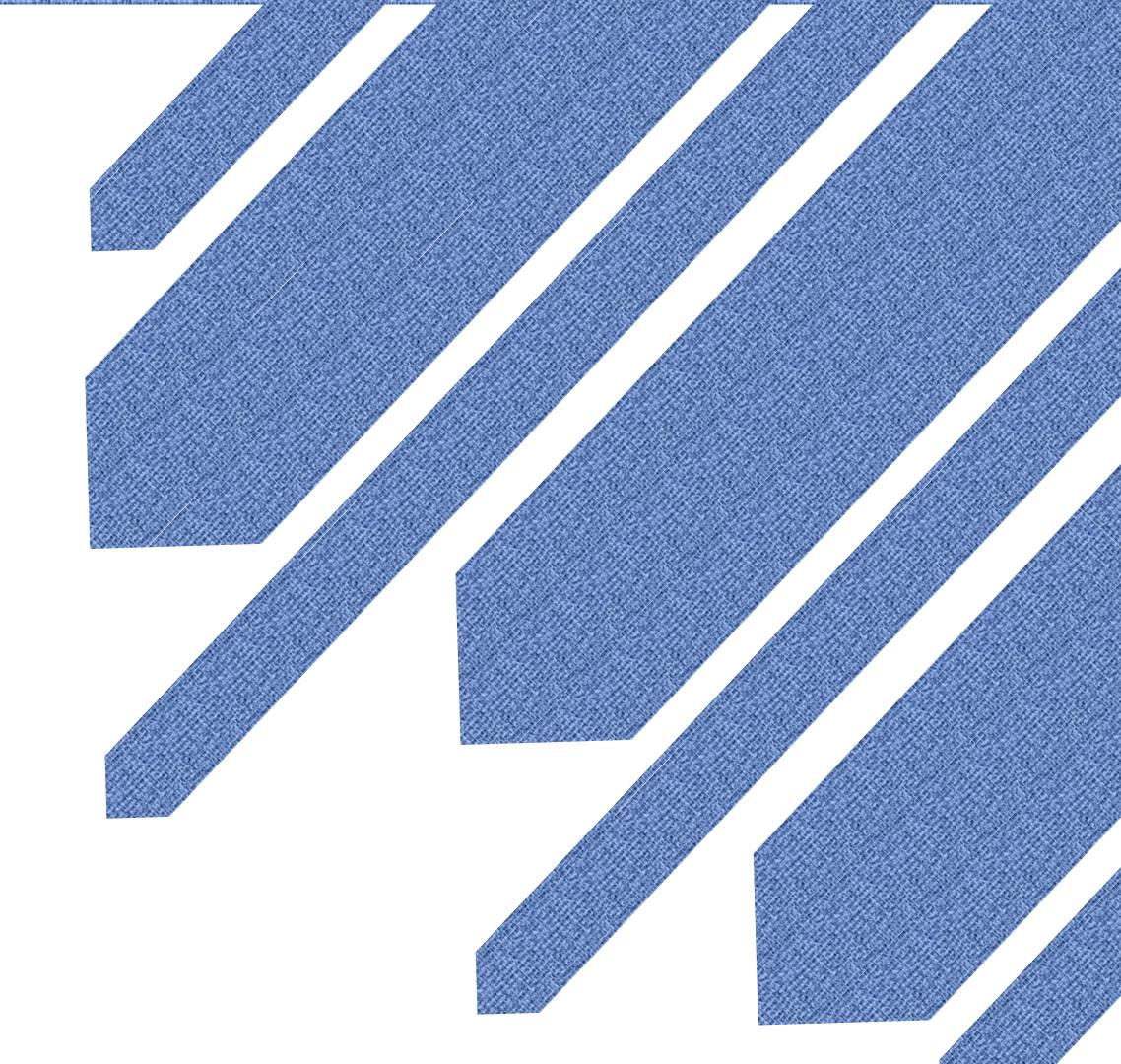
Stopping criteria:

- SSD



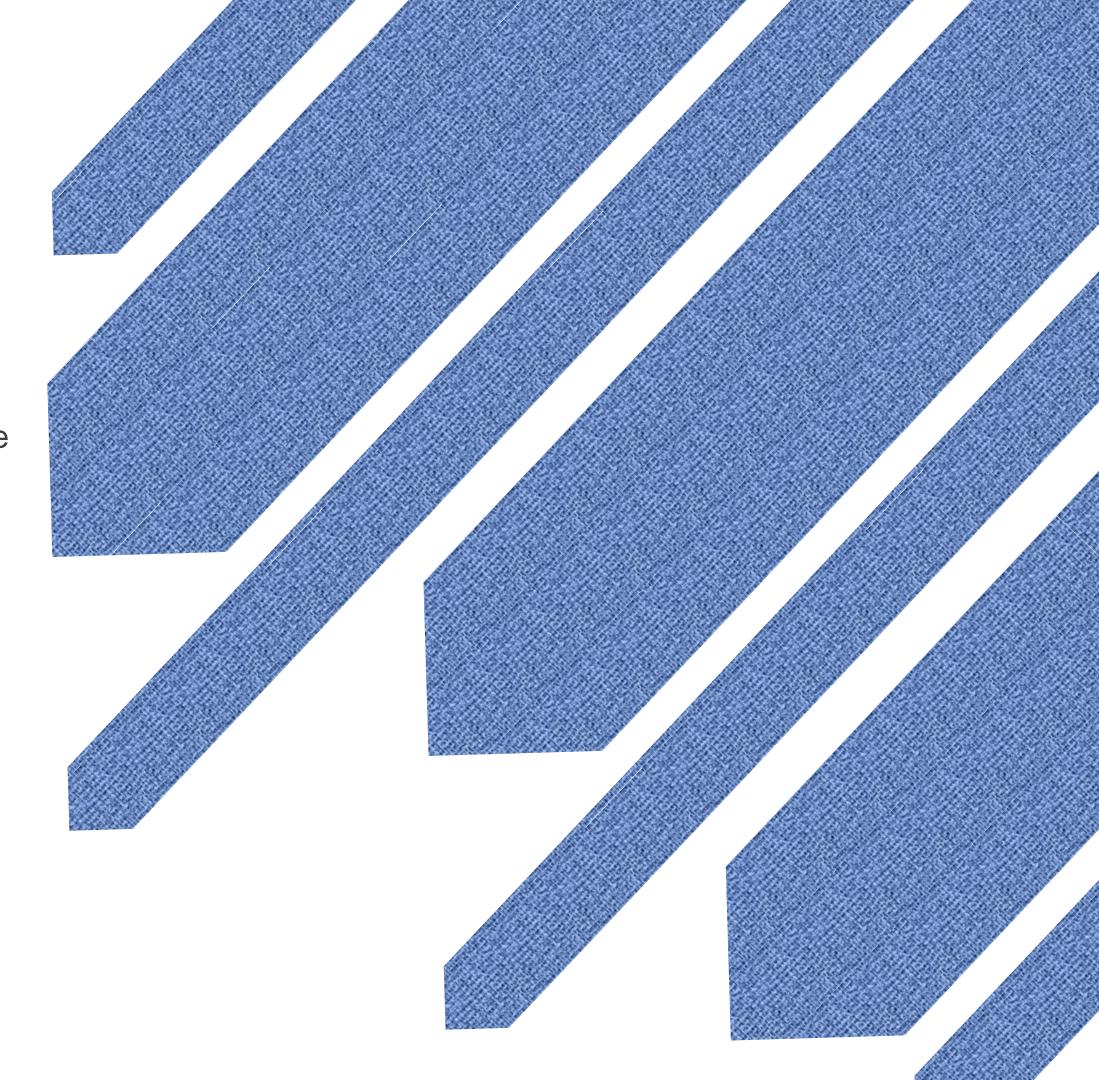
1st Project

World Population Analysis

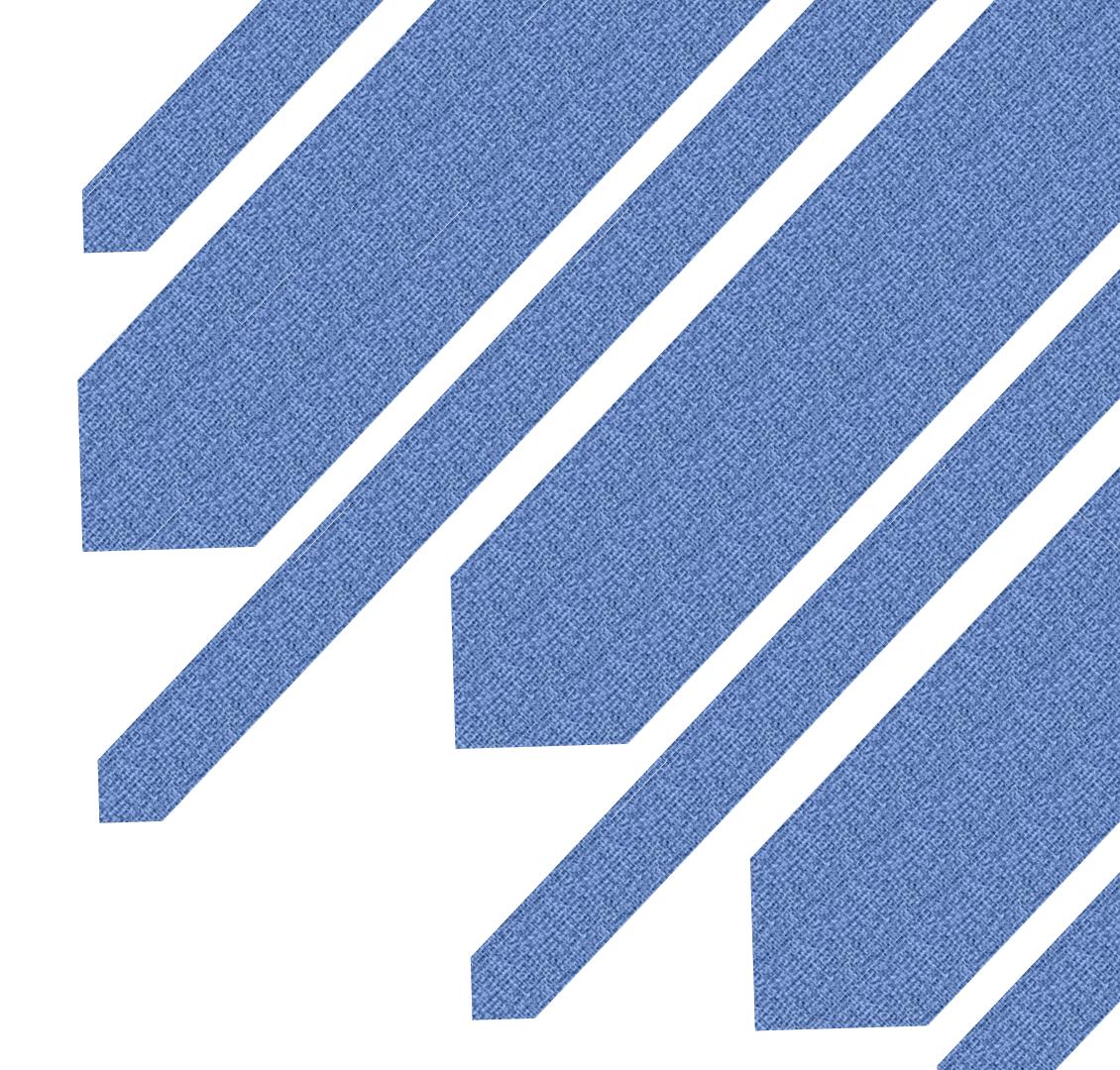


This project analyzed population and land data for various countries and continents over multiple years. It aimed to understand population trends, population growth, and land availability. The findings provide valuable insights into global population dynamics and the distribution of land resources.

- By examining key indicators such as population counts, growth rates, and land area
- we can better understand the dynamics of global population distribution and resource utilization.



Data Understanding

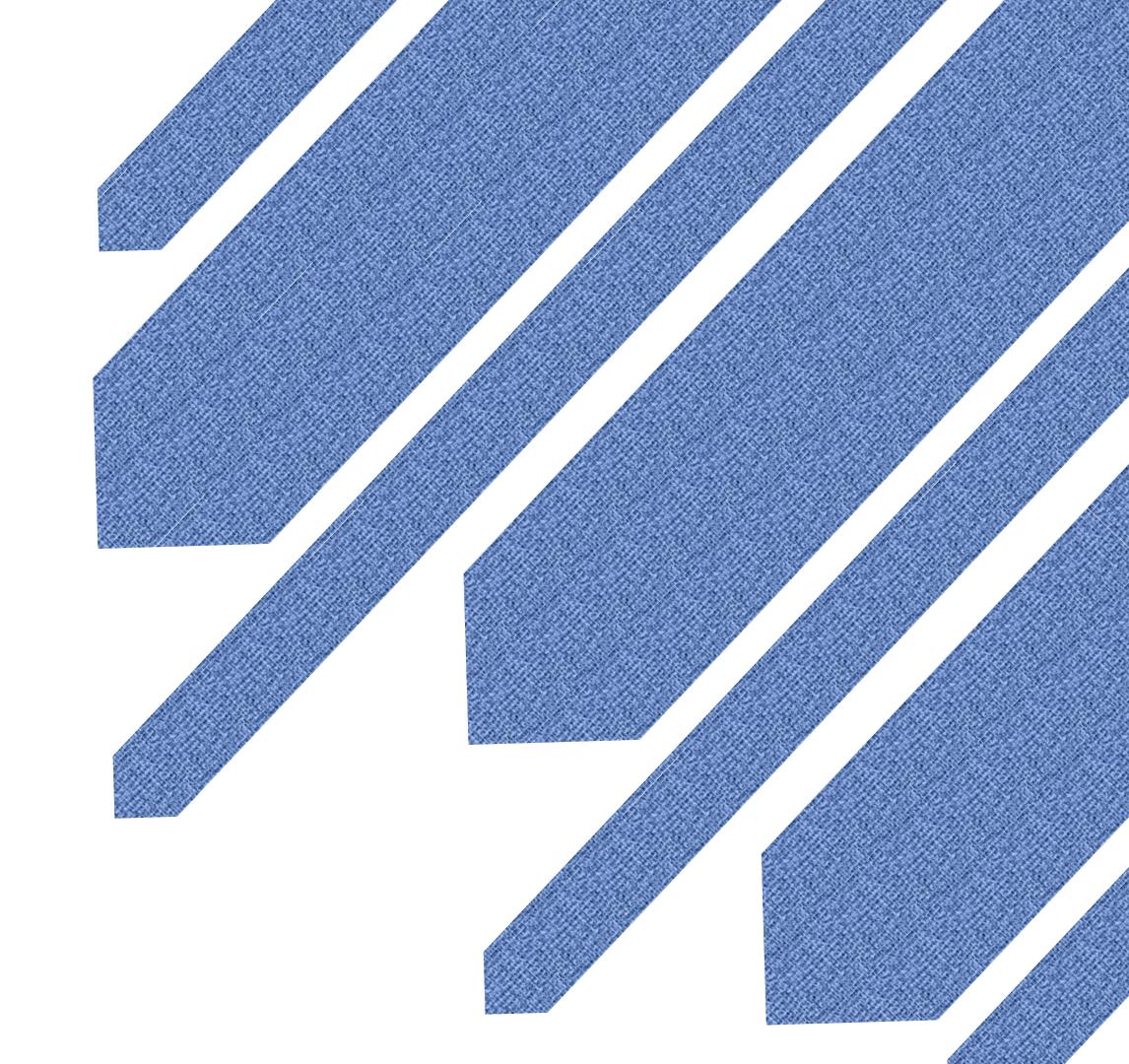


Data Understanding

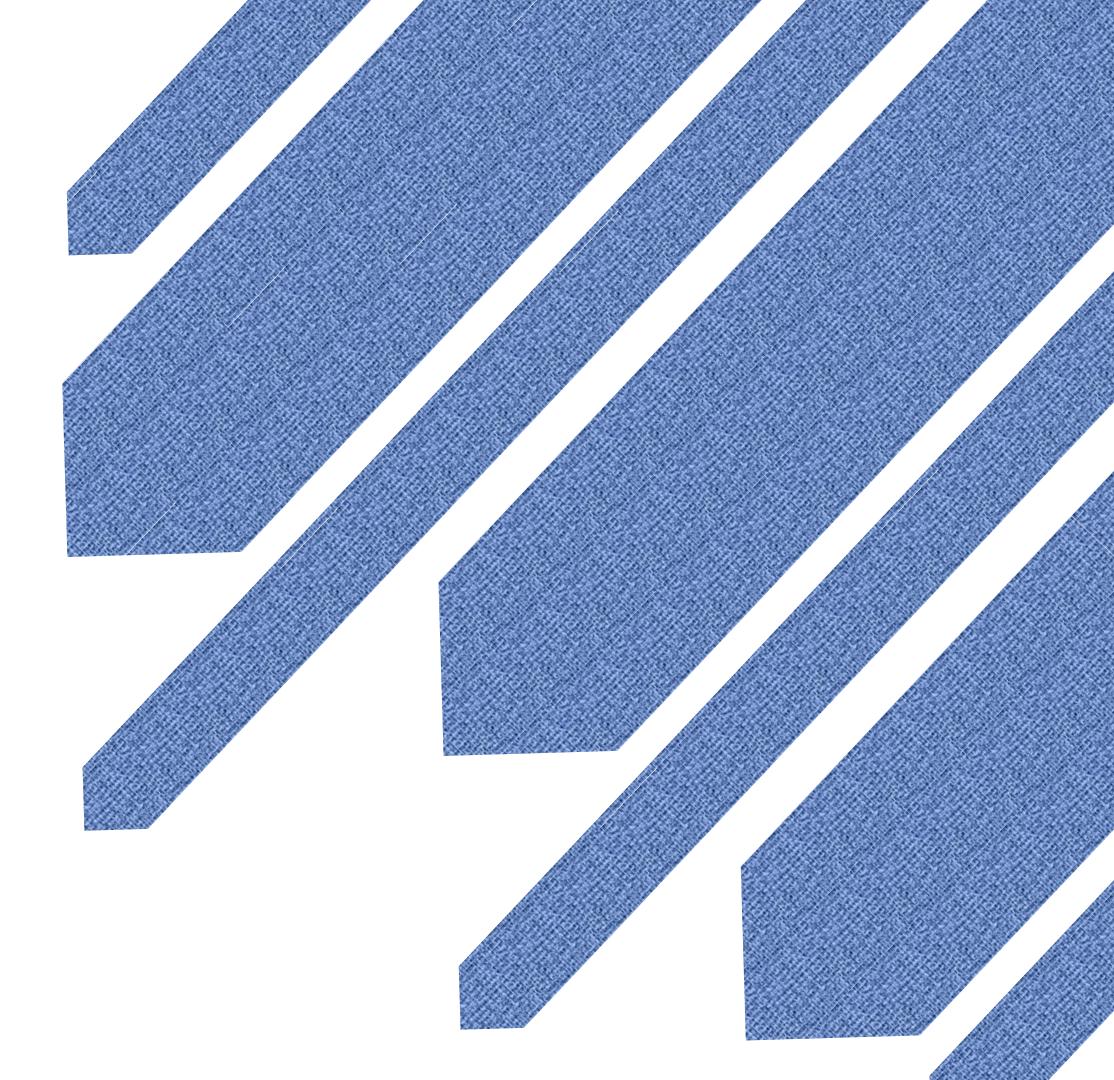
- · We have 15 features.
- We have data of population of Years from 1970 till
 2022 as following (19970, 1980, 1990, 2000, 2010, 2015, 2020, 2022)
- Our data is 234 records and data are clear having no null values.
- Density (per km²): The population density, calculated as the population per square kilometer.
- Growth Rate: The rate of population growth.
- World Population Percentage: The percentage of the world's population represented by the country/territory

Countries per Continent ->>

- Africa 57
- Asia 50
- Europe 50
- North America 40
- Oceania 23
- South America 14

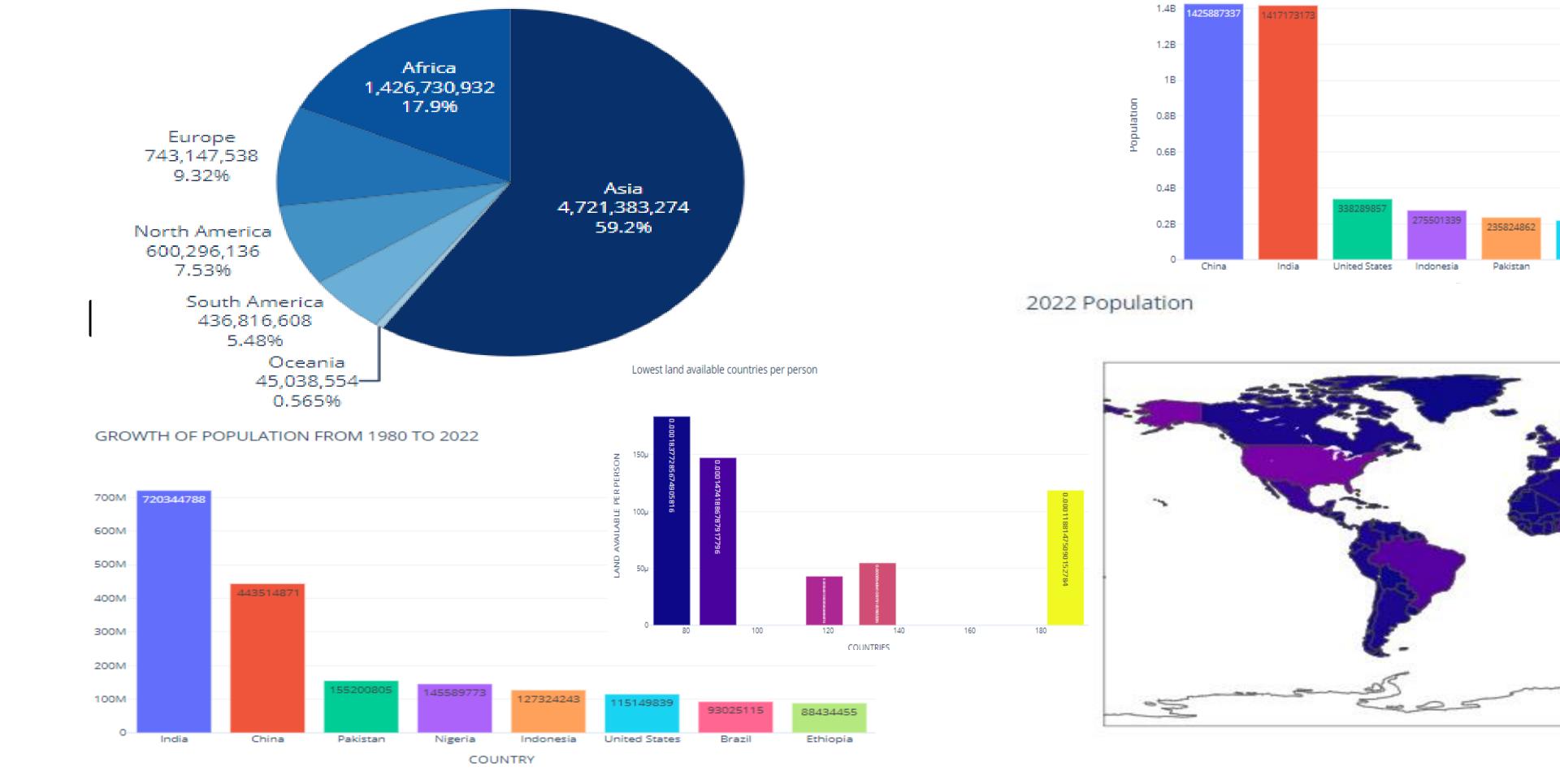


EDA Charts



EDA Charts

2022 Population per Country





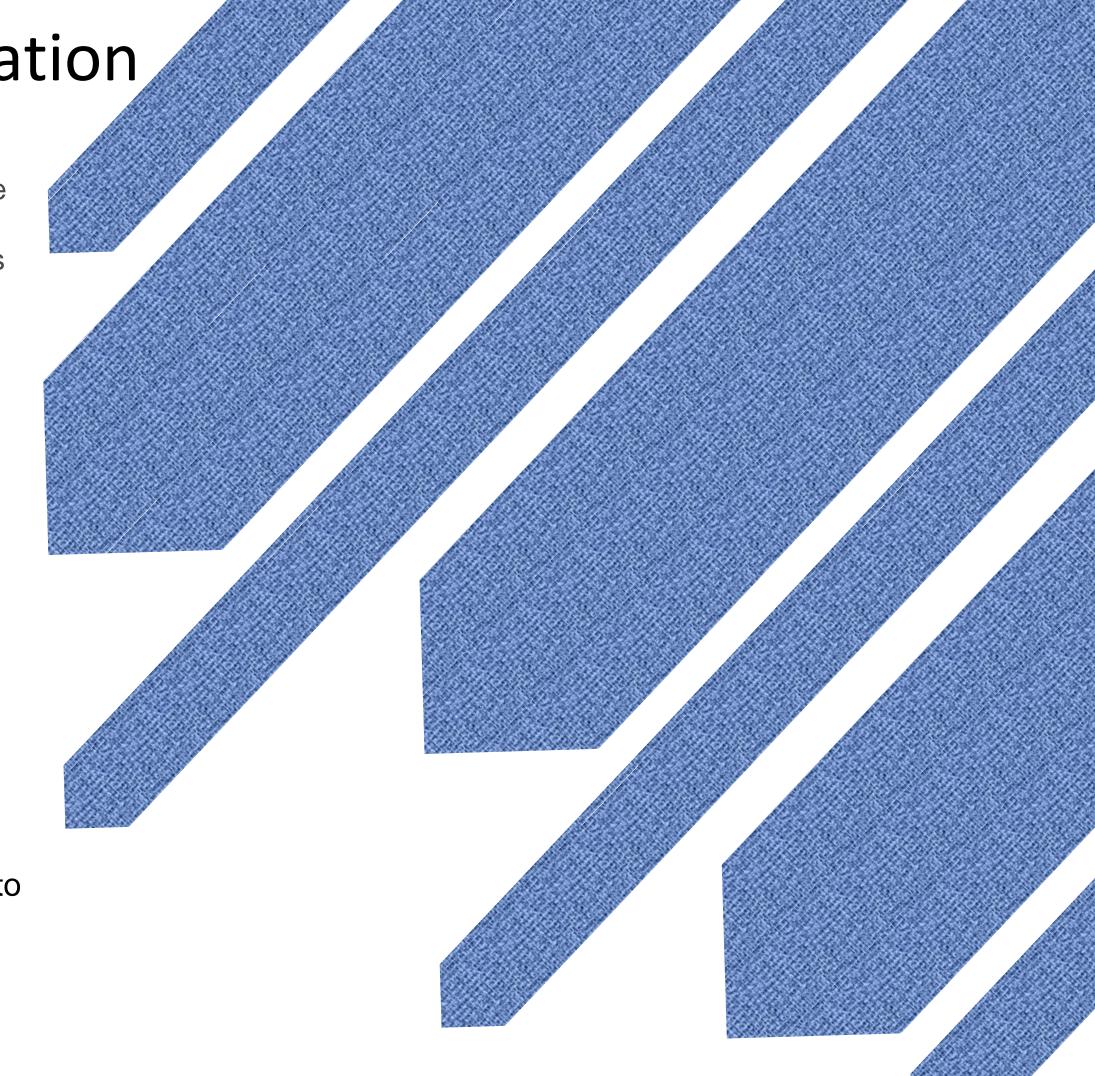
Conclusion & Recommendation

Population Growth: The population of several countries and continents has experienced significant growth over the years, with Asia and Africa showing the highest population increases. This growth poses challenges and opportunities in areas such as infrastructure, healthcare, and resource management.

- Regional Variations: Population distribution varies across continents, with Asia having the largest population overall. The growth rates and patterns differ among continents, influenced by factors such as fertility rates, migration, and economic development.
- Population by Country: Countries like China and India have the highest populations, which impact various aspects of their economies and societies. Understanding population dynamics in these countries is crucial for effective policy-making and resource allocation

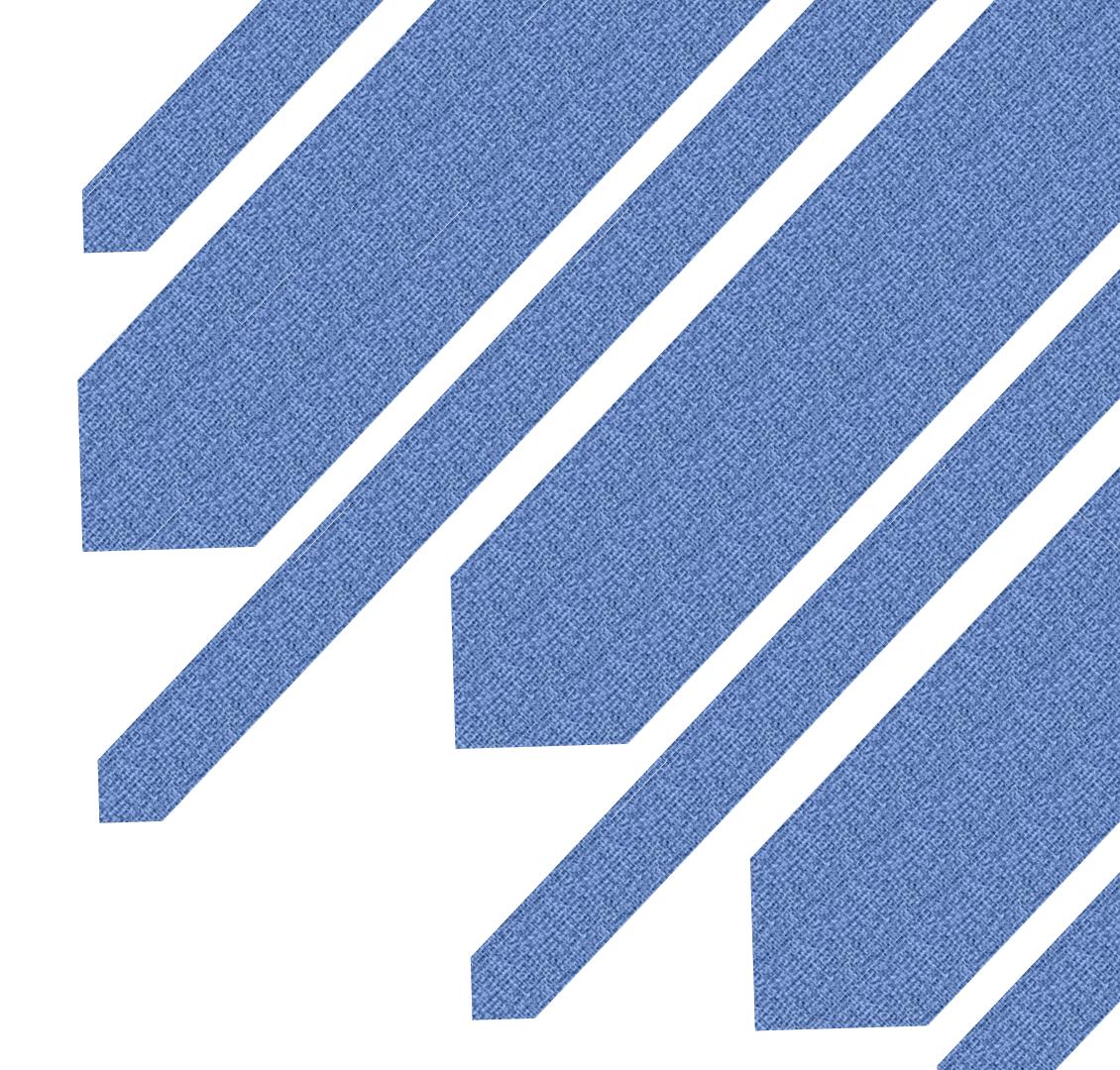
Recommendation

• Implement comprehensive family planning programs to ensure access to reproductive healthcare, education, and contraceptive methods.



2nd Project

Loan Approval Prediction

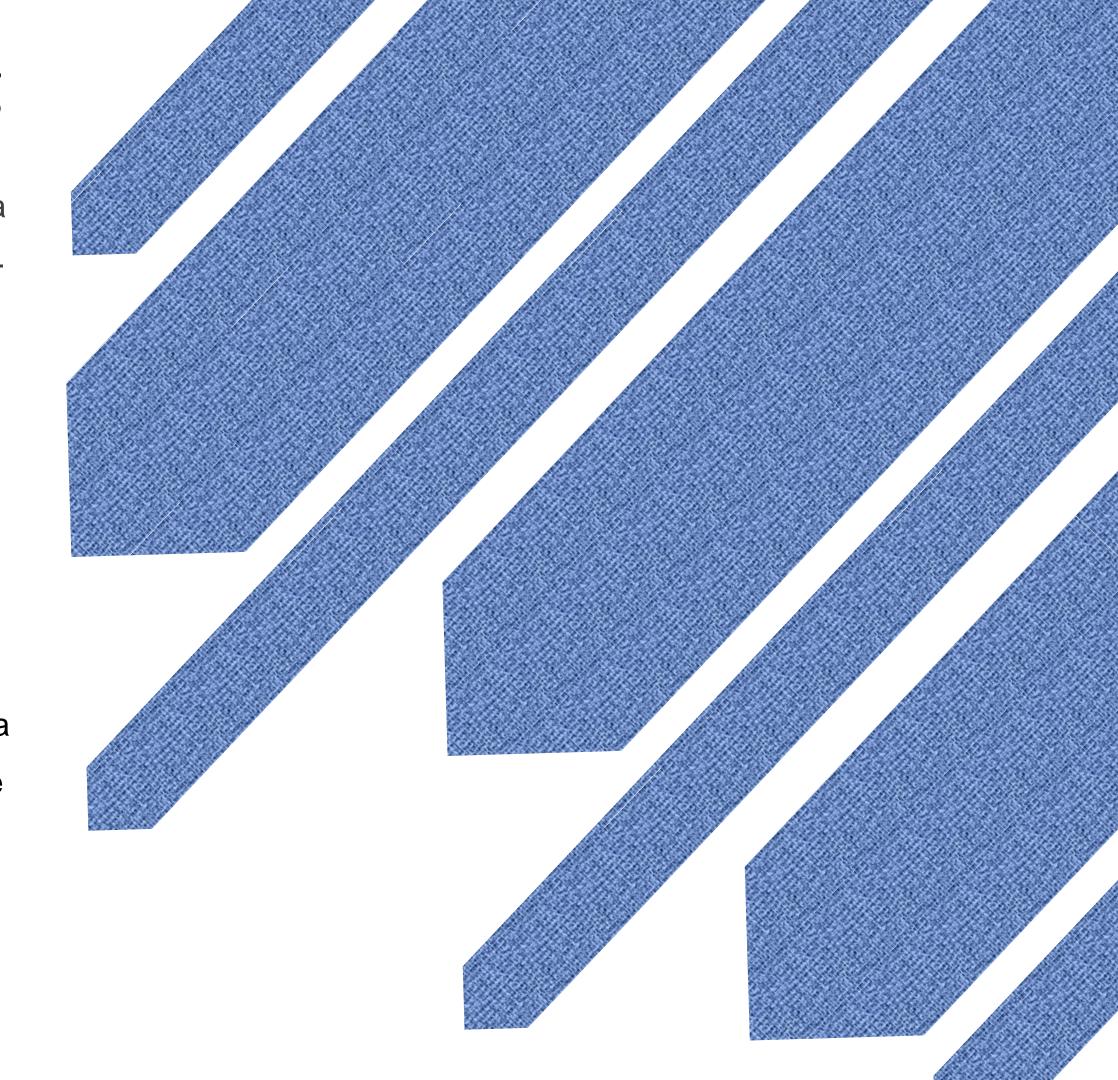


The Loan Approval Prediction project aims to develop a predictive model to assist in the loan approval decision-making process. By analyzing various features of loan applicants, such as credit history, income, education, and property area,

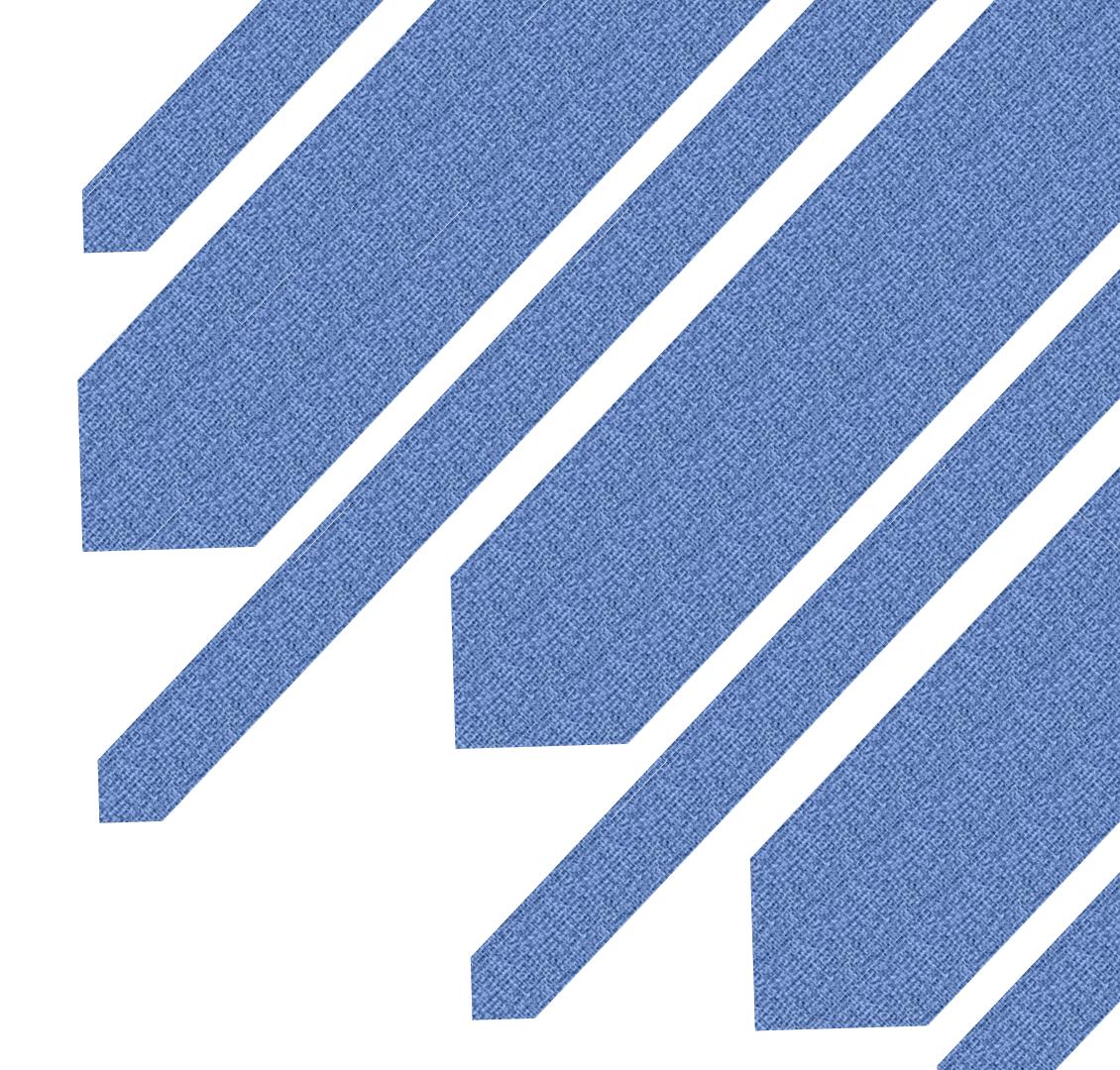
Other Influential Factors: Features such as property area, marital status, education level, and income also play a significant role in loan approval decisions.

Loan Approval Objective:

The main objective is to accurately predict whether a loan application will be approved or not based on the given features. This prediction model can help streamline the loan approval process and make informed decisions.

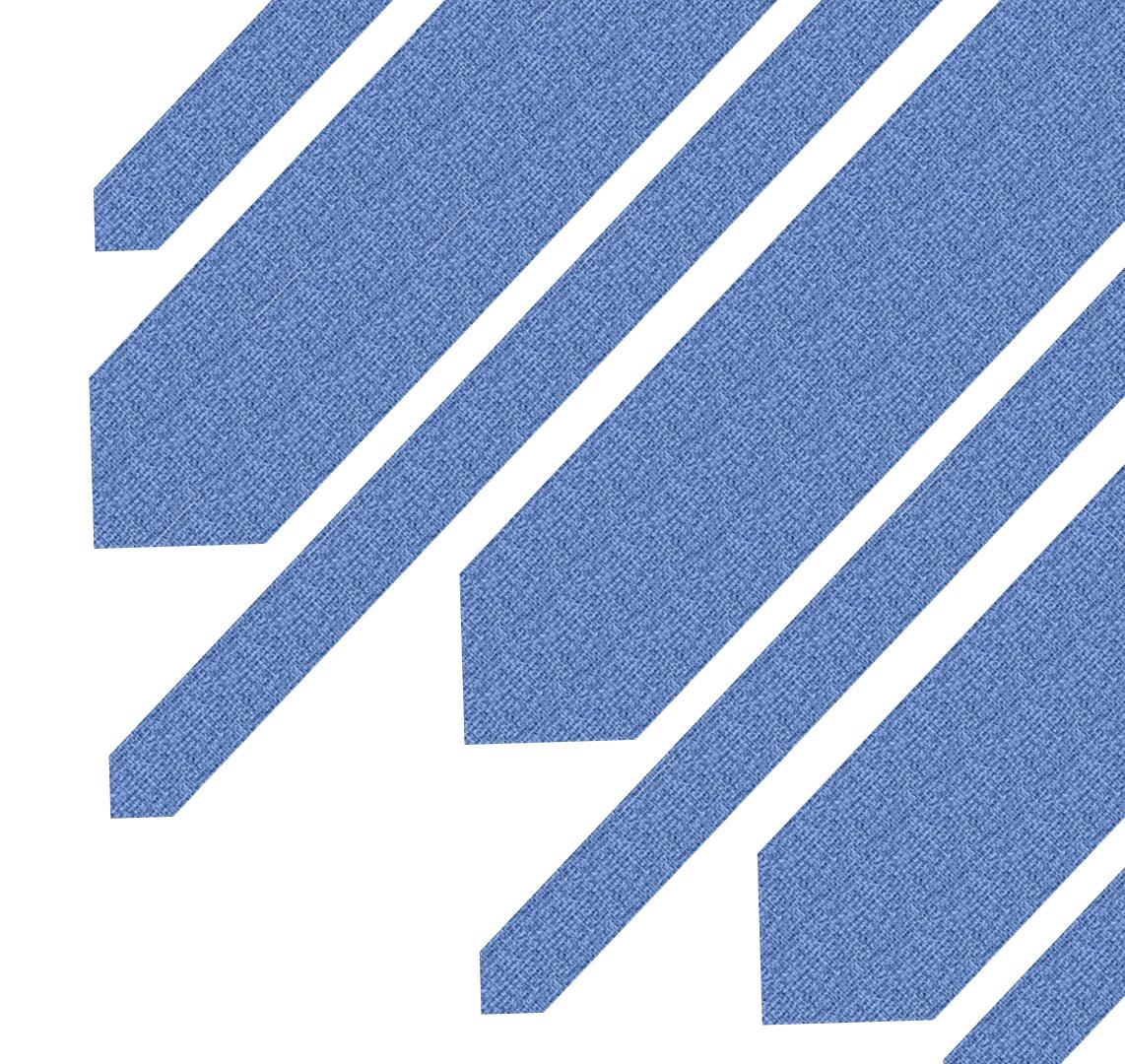


Data Understanding

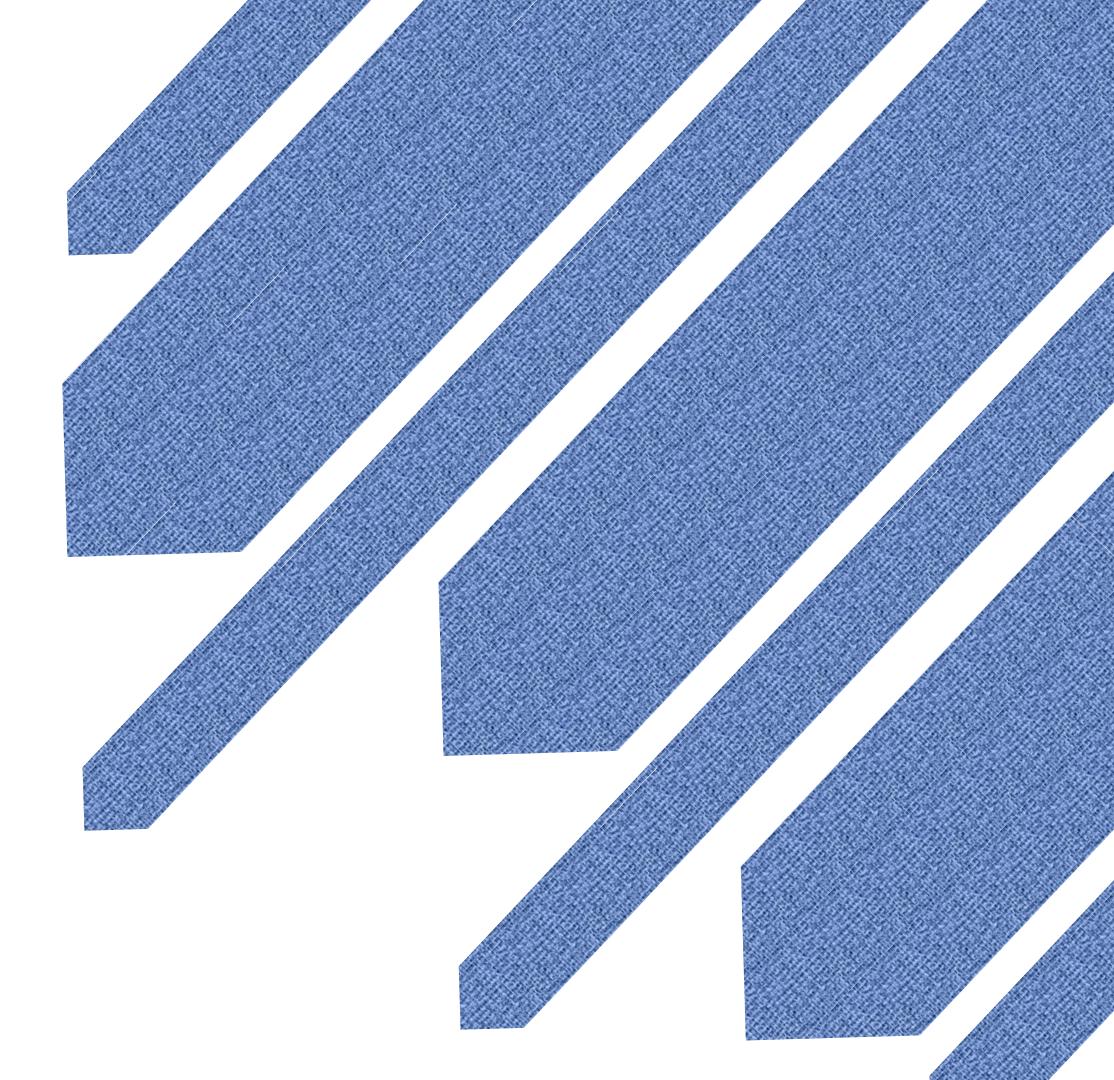


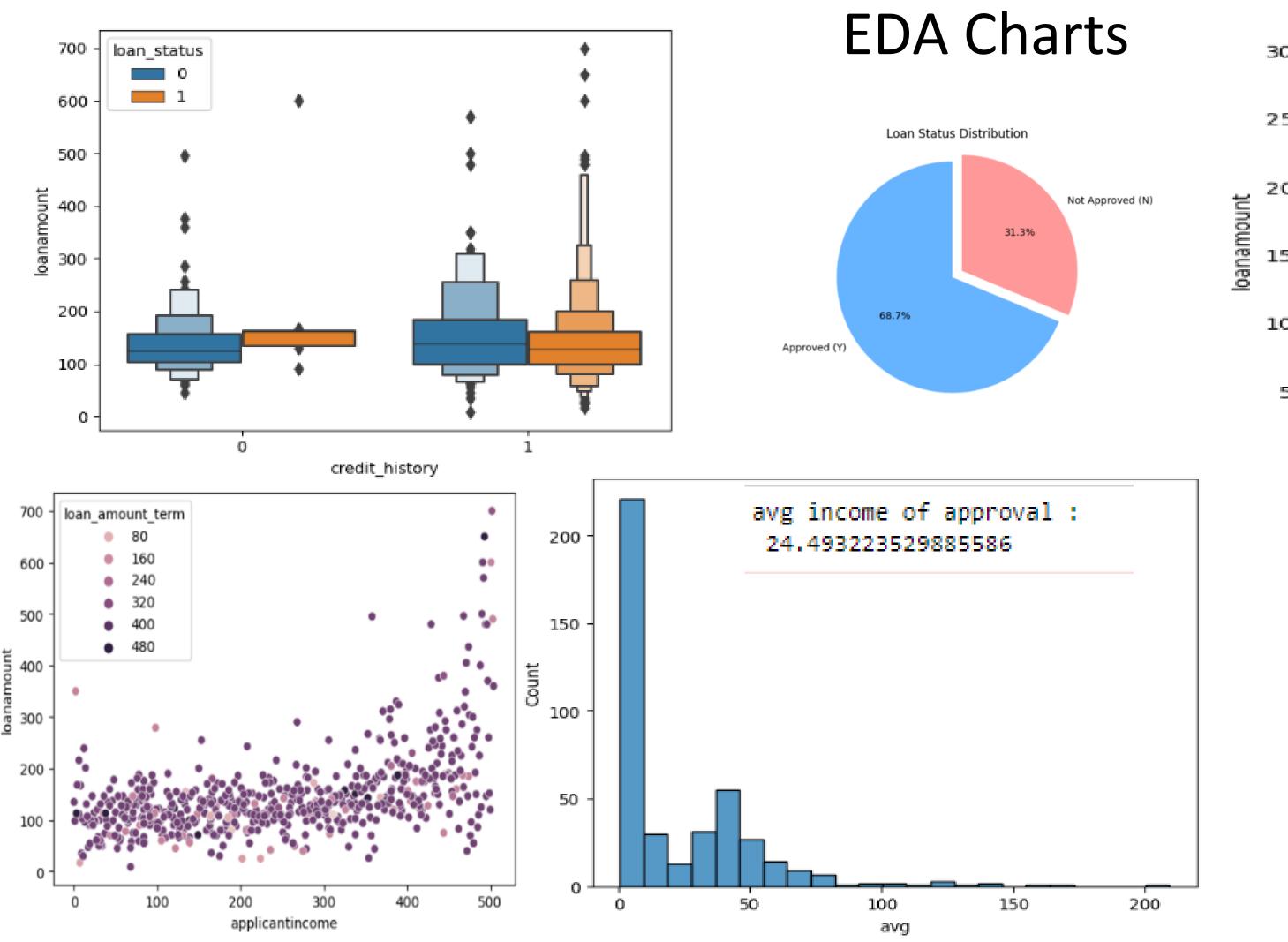
Data Understanding

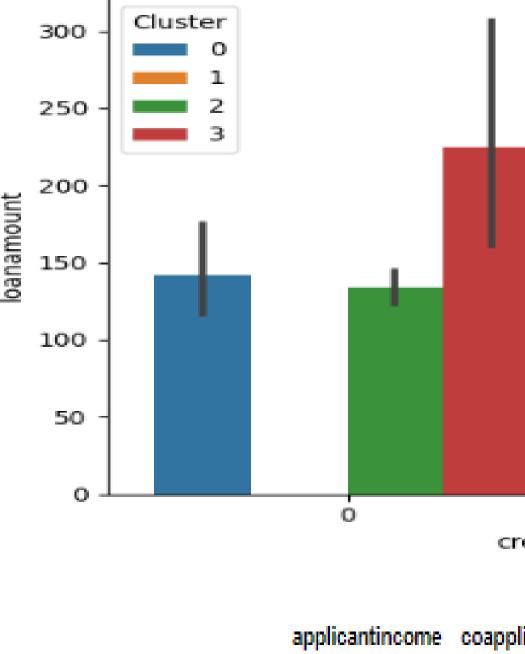
- Column Identification: The dataset consists of the following columns: Loan_ID, Gender, Married, Dependents, Education, Self_Employed, ApplicantIncome, CoapplicantIncome, LoanAmount, Loan_Amount_Term, Credit_History, Property_Area, and Loan_Status.
- Data Exploration: The dataset is explored by analyzing the structure, content, and distribution of each column. This includes checking the data types, unique values, and missing values in the dataset.
- Loan Status Distribution: The distribution of the Loan_Status
 column is examined to understand the proportion of
 approved and not approved loans. This helps in
 understanding the class imbalance and the overall loan
 approval rate in the dataset.
- Missing values are handled by imputing them with appropriate values, and categorical variables are encoded for further analysis.



EDA Charts







	applicantincome	coapp
applicantincome	1.000000	
coapplicantincome	-0.205261	
loanamount	0.497118	
loan_amount_term	-0.026696	

ML Models (If Available)

ML Models (If Available)

Adjusted xgboost

```
from xgboost import XGBClassifier
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score

XGBclassifier = XGBClassifier()

# Define the parameter grid to search
param_grid = {
    'n_estimators': [100, 200, 300], # Number of trees
    'max_depth': [3, 4, 5], # Maximum depth of each tree
    'learning_rate': [0.1, 0.01, 0.001] # Learning rate
}
```

```
# Get the best parameters and best score
best_params = grid_search.best_params_
best_score = grid_search.best_score_

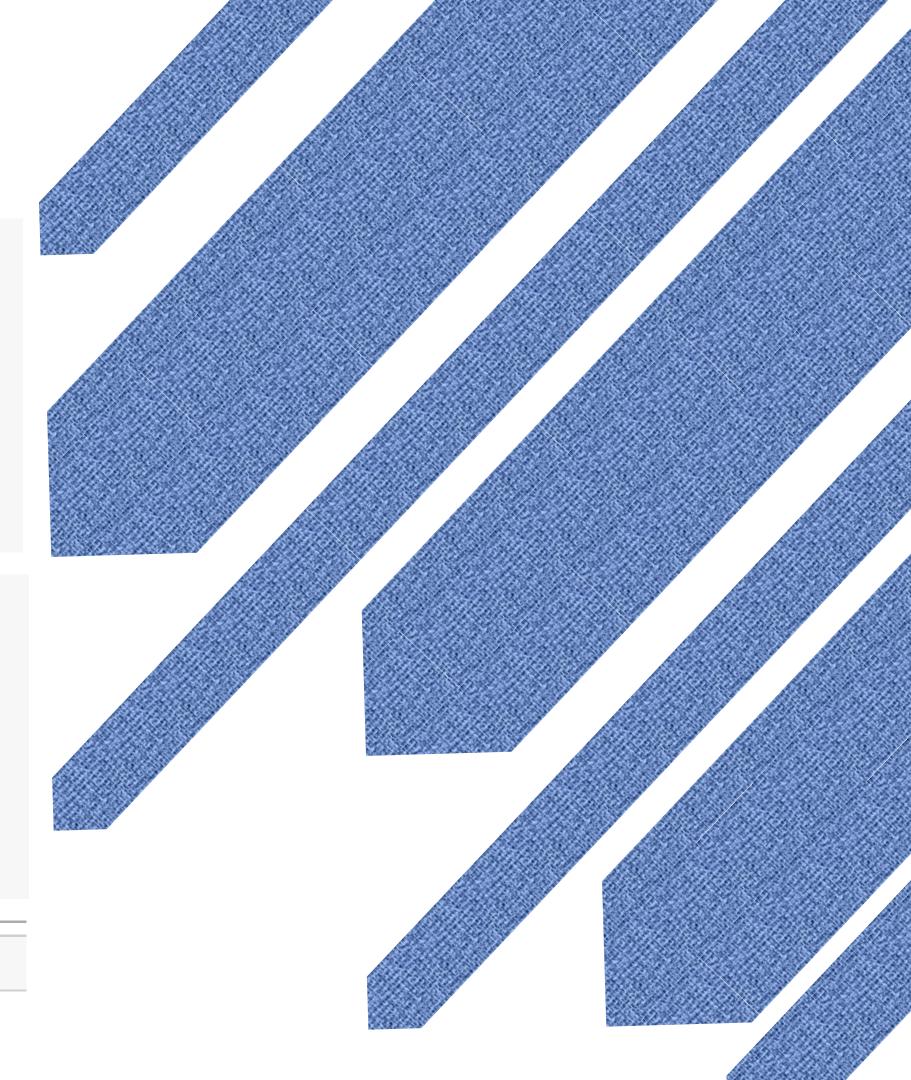
# Create a new XGBoost classifier with the best parameters
XGBclassifier_best = XGBClassifier(**best_params)

# Fit the classifier to the training data
XGBclassifier_best.fit(X_train, y_train)

# Make predictions on the test data
y_pred = XGBclassifier_best.predict(X_test)
```

accuracy

0.827027027027027





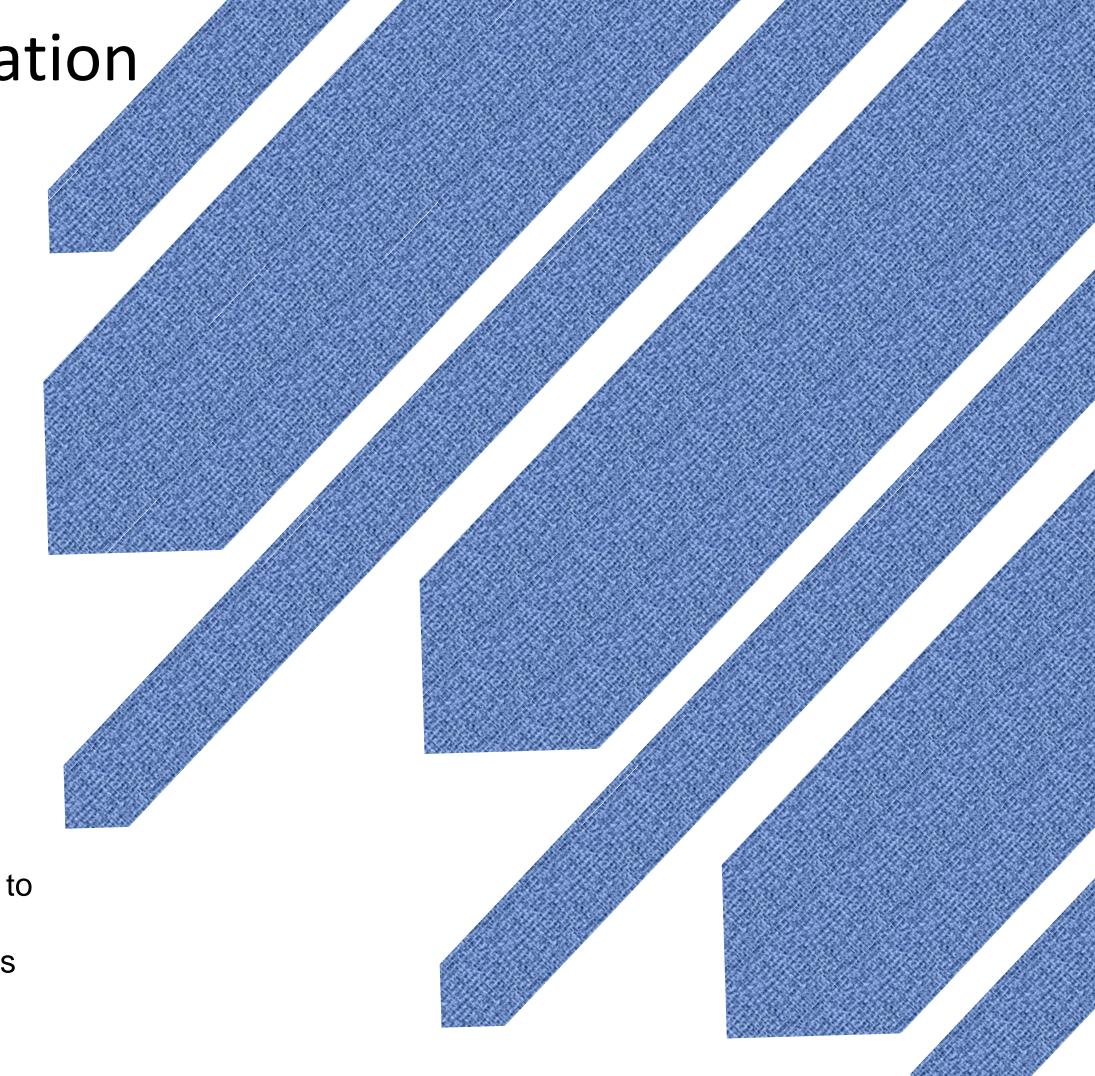
Conclusion & Recommendation

 After preprocessing the data and conducting exploratory data analysis, key insights were derived. Credit history was found to be the most crucial factor in loan approval, followed by property area, marital status, and education. Other factors like income, loan amount, and employment type also played significant roles in loan approval decisions.

- Applicants in semiurban and urban areas had higher loan approval rates compared to those in rural areas.
- Married individuals and graduates had higher loan approval rates.
- Gender and self-employment status had minimal impact on loan approval.
- The models achieved varying levels of accuracy, with Logistic Regression and XGBoost Classifier performing relatively well.
- Adjustments and hyperparameter tuning were applied to improve model performance.

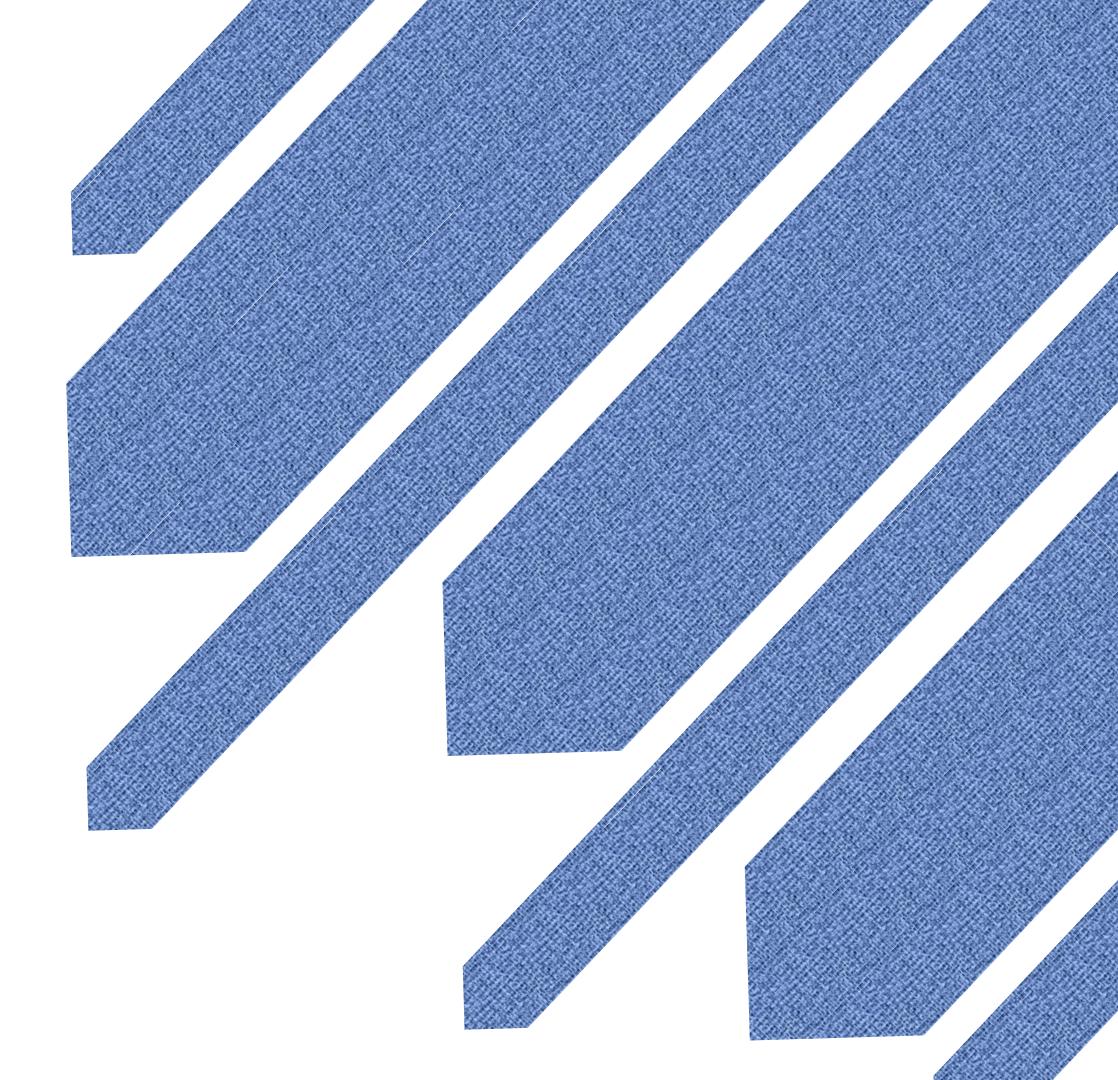
Recommendation

 The project's recommendations include giving priority to applicants with a good credit history, focusing on semiurban and urban areas, considering marital status and education



3nd Project

Sports Wear



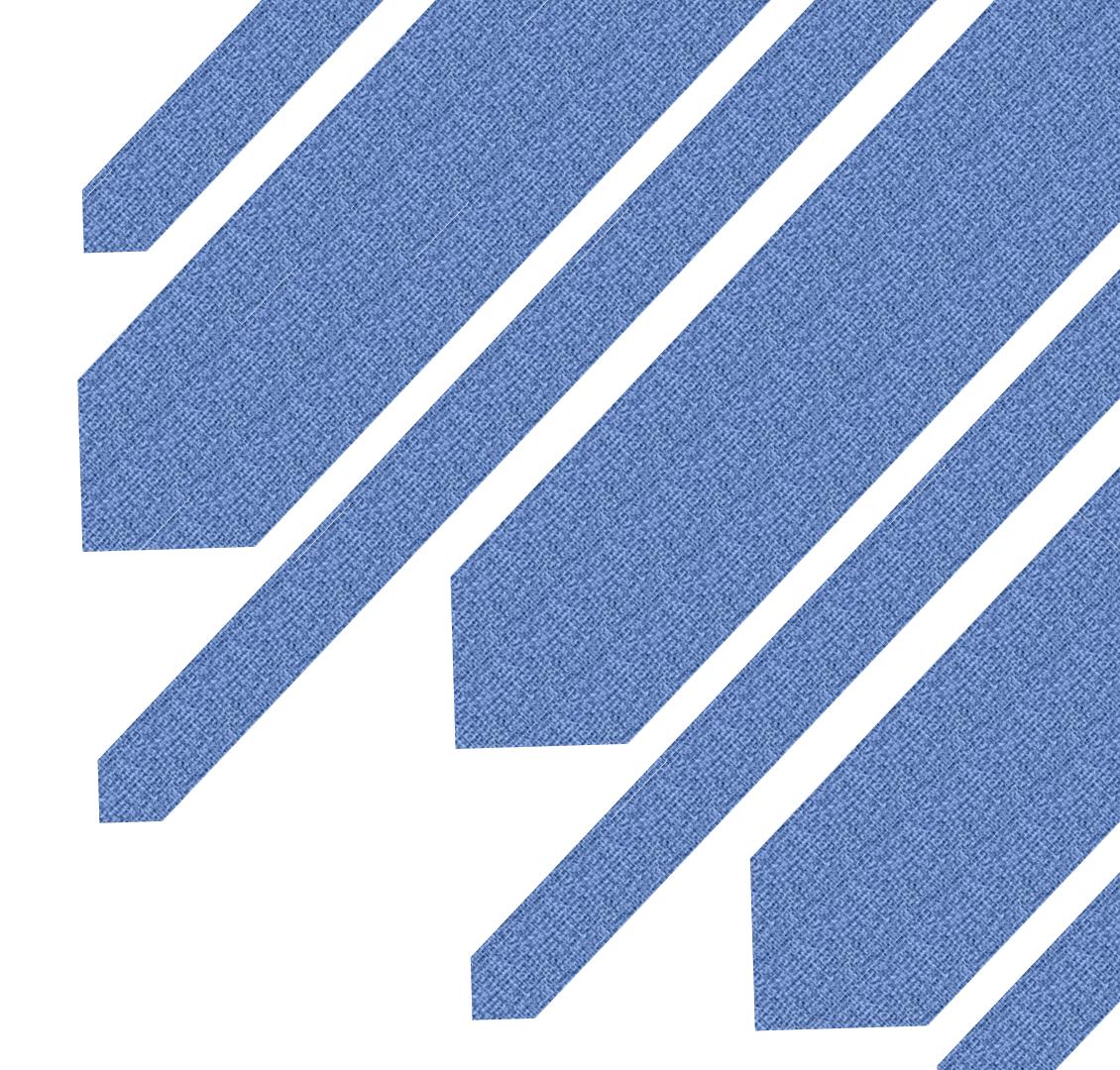
The analysis of the dataset revealed valuable insights into the factors influencing the purchase decision and sales performance of articles. Several key features were examined, including country, article, price, discount, and promotional activities. By understanding the relationship between these features and the purchase decision, businesses can optimize their marketing strategies, pricing, and promotional efforts to drive sales and improve customer satisfaction.

Sports Wear Objective:

 Improve sales and customer satisfaction in the sports wear category. Through the analysis of various factors such as sales performance, customer preferences, pricing, promotions, and market trends, several insights and recommendations were generated to achieve the objective.



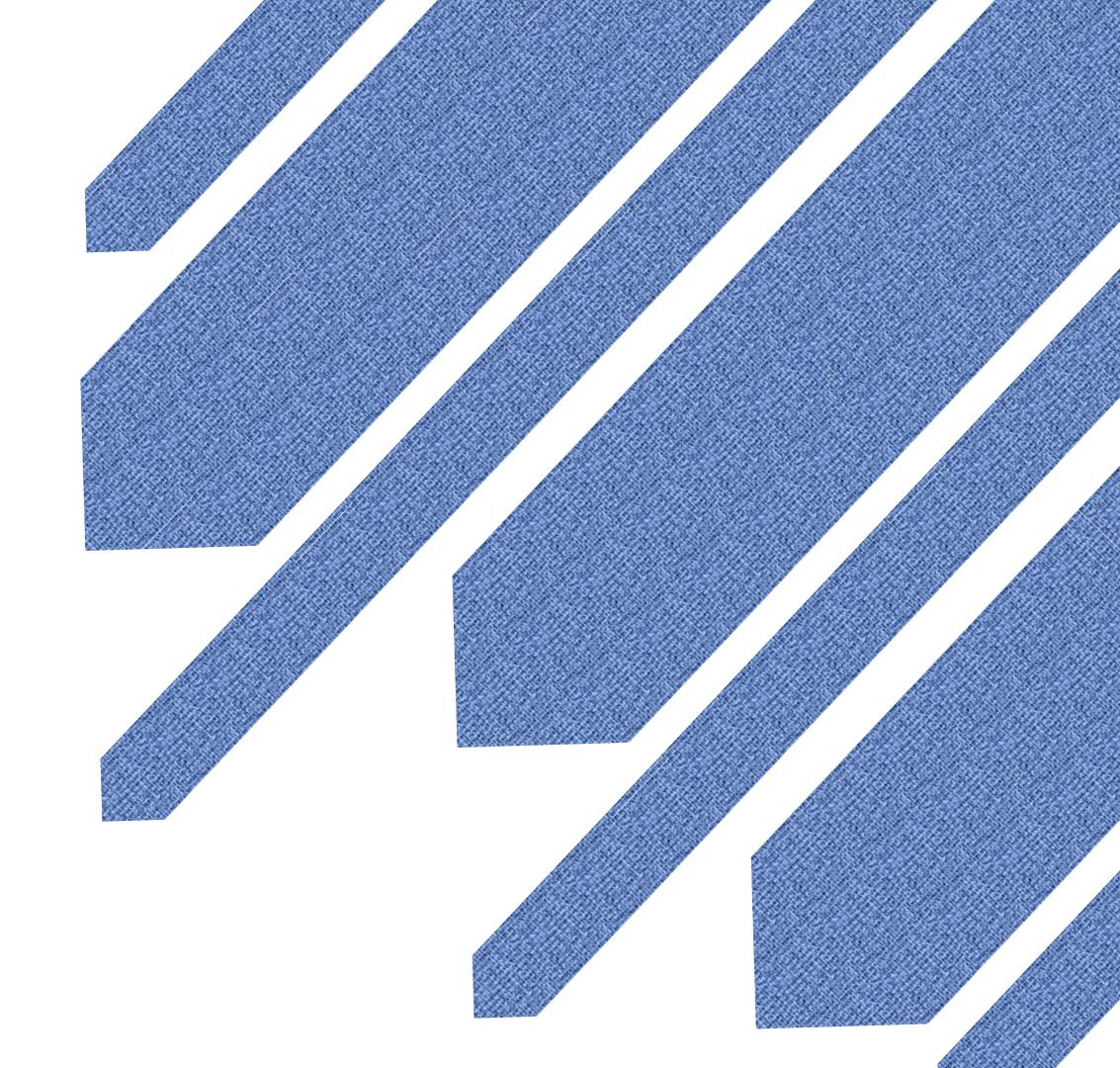
Data Understanding



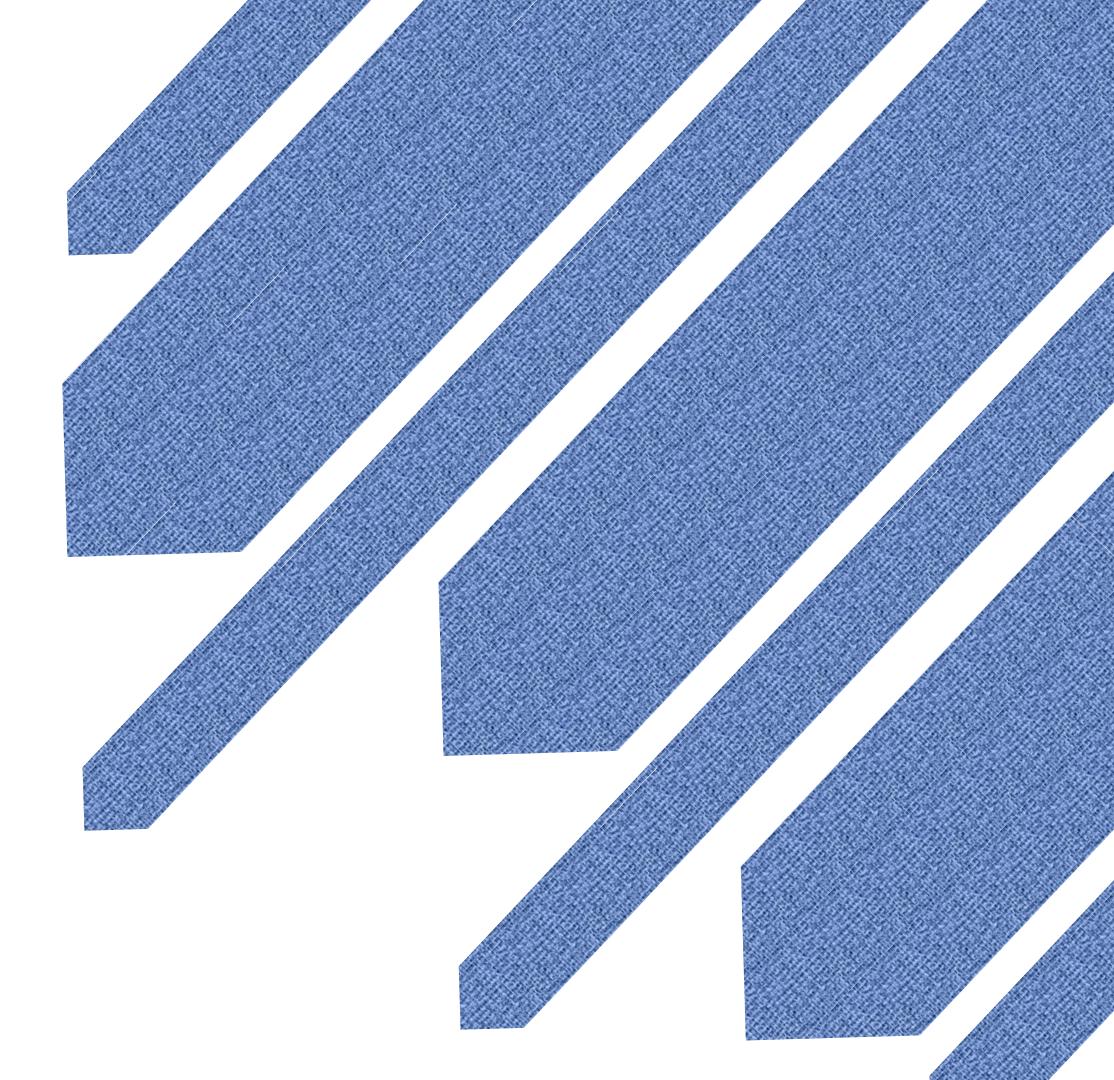
(100000, 24)

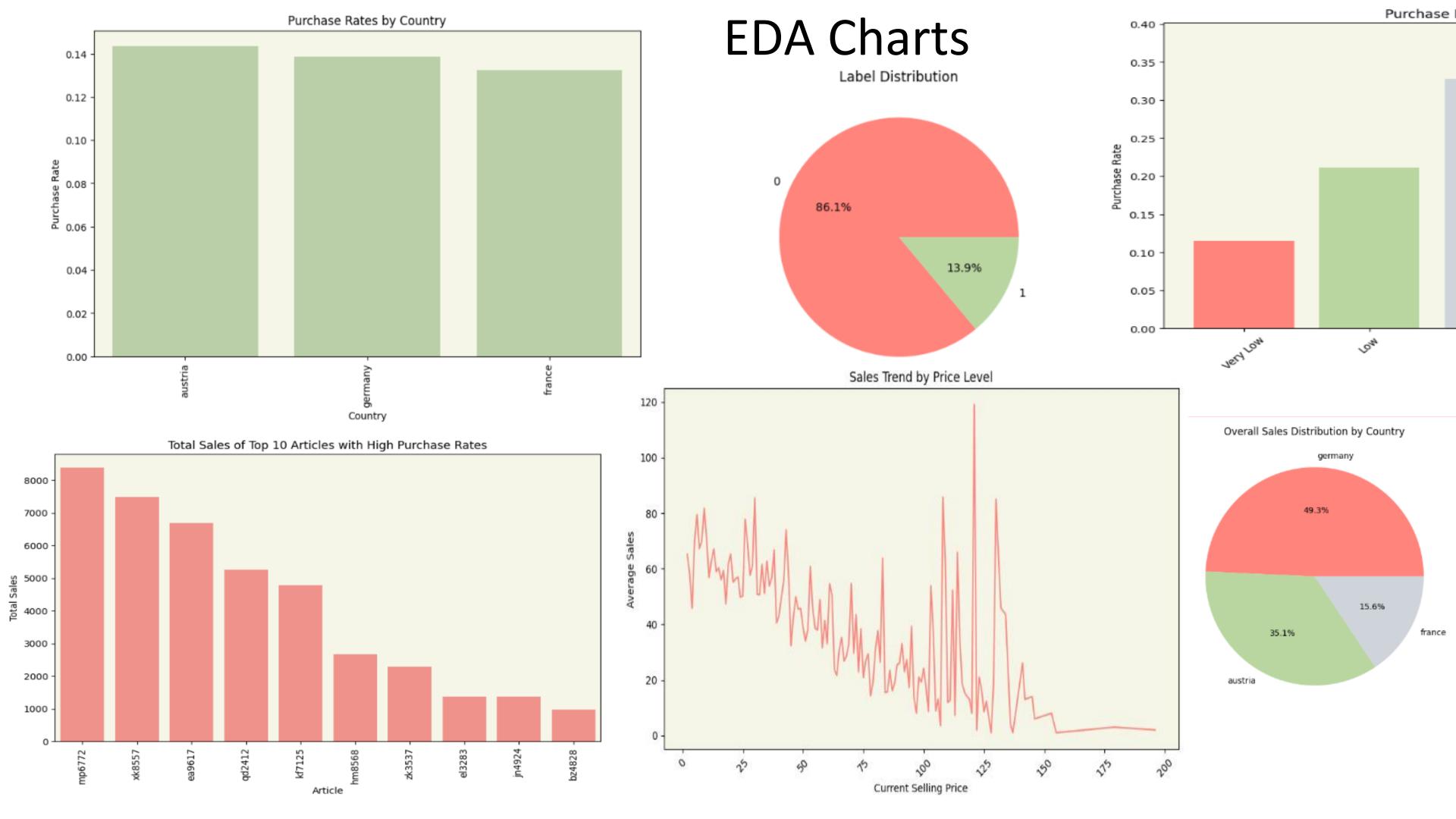
Data Understanding

- Gather the dataset related to sports wear, including information about articles, prices, promotions, sales, customer demographics, and other relevant variables.
- Our dataset contains 10000 records and 24 Features .
- label Distribution: The label distribution is imbalanced, with Label 0 representing approximately 86.1% of the dataset, while Label 1 represents around 13.9% of the dataset.
- This indicates a significant class imbalance, with Label 0 being the majority class and Label 1 being the minority class. Imbalanced data can impact the performance of machine learning models, as they may have a tendency to favor the majority class, potentially leading to biased predictions and lower accuracy for the minority class.



EDA Charts





ML Models (If Available)



Conclusion & Recommendation

1. Country Influence: Austria, Germany, and France showed higher purchase rates and sales performance. Businesses should focus on these countries for targeted marketing and expansion opportunities.

2. Article Popularity: Certain articles, such as "XK8557" and "MP6772," showed higher purchase rates and total sales. Promoting these top-performing articles can maximize profitability.

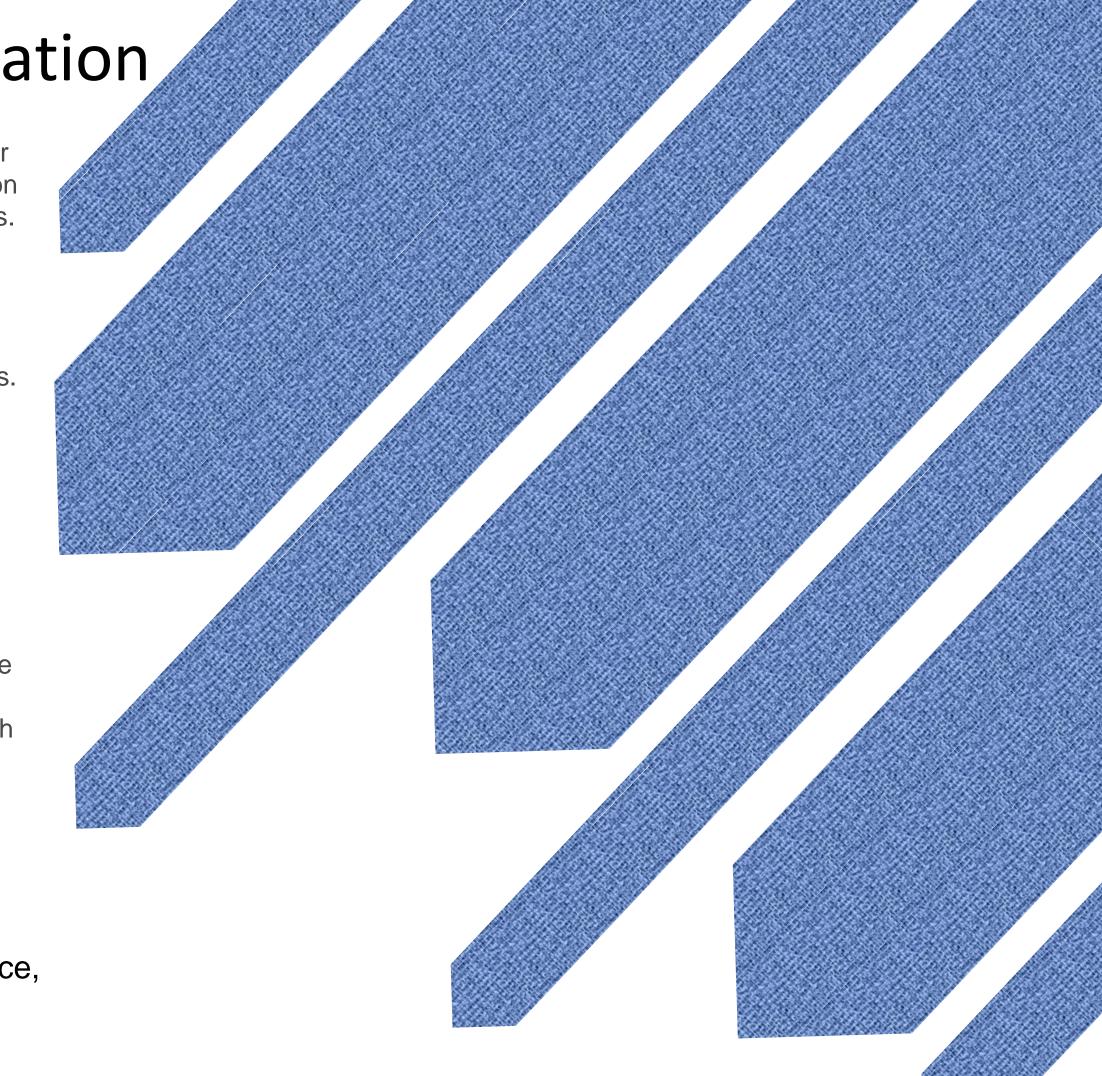
3. Price Impact: The regular price of articles did not show a significant difference between purchased and non-purchased items. However, analyzing the current selling price revealed that lower prices and moderate discounts had a positive correlation with higher purchase rates.

4. Discount Effect: High discounts were not associated with higher purchase rates. Instead, articles with low to medium discounts showed better sales performance. Businesses should focus on providing moderate discounts that strike a balance between attractiveness and maintaining profit margins.

5. Promotional Activities: Media advertisements (promo1) and store events (promo2) had a positive impact on the purchase decision and sales. Coordinating different promotions and aligning them with price ranges can maximize their effectiveness in driving sales.

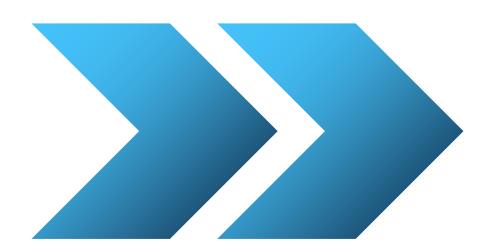
Recommendation

 Targeted Marketing: Focus on countries with higher purchase rates, such as Austria, Germany, and France, to maximize marketing efforts and expand customer base.









I am grateful for the time you invested in us, constructive feedback on my projects, or simply taking the time to answer my questions. Your willingness to share your experiences and provide guidance, even amidst your busy schedule, has made a lasting impression on me.

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