

A COMPREHENSIVE MEASURE OF WELL-BEING : THE HUMAN DEVELOPMENT INDEX

AN INDUSTRY ORIENTED MINI REPORT

Submitted to

JAWAHARLAL NEHRU TECHNOLOGICAL UNIVERSITY, HYDERABAD

In partial fulfillment of the requirements for the award of the degree of

BACHELOR OF TECHNOLOGY

In

COMPUTER SCIENCE AND ENGINEERING(CSE)

Submitted By

DEEKSHITH PATTIPAKA

SYED IRSHAD AHMED

SATHWIKA PAKALA

UDAYKRISHNA YALAMANCHILI

22UK5A0515

21UK1A05D4

21UK1A05F0

21UK1A05E6

Under the guidance of

MRS. A. SWATHI

Assistant Professor



**DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING
VAAGDEVI ENGINEERING COLLEGE**

Affiliated to JNTUH, HYDERABAD

BOLLIKUNTA, WARANGAL (T.S) – 506005

DEPARTMENT OF

DEPARTMENT OF
COMPUTER SCIENCE AND ENGINEERING(CSE)
VAAGDEVI ENGINEERING COLLEGE(WARANGAL)



CERTIFICATE OF COMPLETION
INDUSTRY ORIENTED MINI PROJECT

This is to certify that the UG Project Phase-1 entitled “A COMPREHENSIVE MEASURE OF WELL-BEING: THE HUMAN DEVELOPMENT INDEX USING MACHINE LEARNING” is being submitted by DEEKSHITH PATTIPAKA(22UK15A0515),SATHWIKAKAKALA(21UK1A05F0),SYED IRSHADAHMED(21UK1A05D4),UDAYKRISHNAYALAMANCHILI(21UK1A05E6) in partial fulfillment of the requirements for the award of the degree of Bachelor of Technology in Computer Science & Engineering to Jawaharlal Nehru Technological University Hyderabad during the academic year 2023- 2024.

Project Guide

Mrs.A.Swathi

(Assistant Professor)

HOD

Dr.R.Naveen Kumar

(Professor)

External

ACKNOWLEDGEMENT

We wish to take this opportunity to express our sincere gratitude and deep sense of respect to our beloved **Dr.P.PRASAD RAO**, Principal, Vaagdevi Engineering College for making us available all the required assistance and for his support and inspiration to carry out this UG Project Phase-1 in the institute.

We extend our heartfelt thanks to **Dr.R.NAVEEN KUMAR**, Head of the Department of CSE, Vaagdevi Engineering College for providing us necessary infrastructure and thereby giving us freedom to carry out the UG Project Phase-1.

We express heartfelt thanks to Smart Bridge Educational Services Private Limited, for their constant supervision as well as for providing necessary information regarding the UG Project Phase-1 and for their support in completing the UG Project Phase-1.

We express heartfelt thanks to the guide, **A.SWATHI**, Assistant professor, Department of CSE for his constant support and giving necessary guidance for completion of this UG Project Phase-1.

Finally, we express our sincere thanks and gratitude to my family members, friends for their encouragement and outpouring their knowledge and experience throughout the thesis.

DEEKSHITH PATTIPAKA

SYED IRSHAD AHMED

SATHWIKA PAKALA

UDAYKRISHNA YALAMANCHILI

22UK5A0515

21UK1A05D4

21UK1A05F0

21UK1A05E6

ABSTRACT

The Human Development Index (HDI) is a composite statistic used to rank countries based on human development levels, traditionally measured through life expectancy, education, and per capita income indicators. Despite its widespread use, the HDI has limitations in capturing the complexity and nuances of human well-being. This project aims to enhance the HDI by leveraging machine learning techniques to provide a more comprehensive and accurate measure of human development.

In this study, we develop a predictive model using machine learning algorithms to estimate the HDI based on a broader set of features. The model incorporates additional variables such as mean years of schooling, gross national income per capita, and other relevant socio-economic indicators. By analyzing data from various countries, the model identifies patterns and relationships that traditional methods may overlook.

We implemented a Flask web application to make the predictive model accessible and user-friendly. The application allows users to input relevant data, including life expectancy, education levels, income, and country-specific information. The model then processes this data and provides an HDI prediction, categorizing it into low, medium, high, or very high HDI.

The results demonstrate that the machine learning-enhanced HDI offers a more nuanced understanding of human development, capturing variations that the traditional HDI might miss. This approach not only improves the accuracy of HDI predictions but also provides policymakers and researchers with deeper insights into the factors influencing human well-being.

Overall, this project highlights the potential of machine learning in redefining traditional metrics and advancing our understanding of global development. By providing a more detailed and dynamic measure of HDI, we aim to support more effective policy-making and contribute to the broader discourse on human development.

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1.INTRODUCTION

1.1.OVERVIEW

The project titled "A Comprehensive Measure Of Well-Being Human Development Index Using Machine Learning" aims to improve the accuracy and comprehensiveness of the Human Development Index (HDI) by employing machine learning techniques and incorporating additional socio-economic indicators. The HDI, a widely used metric for evaluating countries' development levels, traditionally relies on life expectancy, education, and income. However, it has limitations in fully capturing the complexities of human well-being. This project seeks to address these limitations by leveraging machine learning to provide a more nuanced measure of HDI. The methodology involves collecting data from international databases, including variables such as life expectancy, mean years of schooling, GNI per capita, and other socio-economic indicators. Machine learning models are trained to predict HDI using this data, with efforts made to evaluate and optimize the models for accurate predictions. A user-friendly web application is developed using Flask to allow users to input data and receive HDI predictions, displaying the predicted HDI and categorizing it into low, medium, high, or very high HDI. The expected outcome is a more accurate and comprehensive HDI measure, accessible through the web application, providing deeper insights into human development and aiding policymakers in making informed decisions. Ultimately, this project enhances the traditional HDI, offering a better understanding of global development and supporting more effective policy-making to improve human well-being.

1.2.PURPOSE

The purpose of the "Enhancing the Human Development Index Using Machine Learning" project is to advance the traditional Human Development Index (HDI) by integrating machine learning techniques and additional socio-economic indicators. The project aims to address the limitations of the traditional HDI, which relies on basic metrics like life expectancy, education, and income, but fails to capture the full complexity of human well-being. By employing machine learning, the project seeks to provide a more accurate, nuanced, and comprehensive measure of human development. This enhanced HDI will offer deeper insights into the factors affecting human well-being, support more informed decision-making by policymakers, and contribute to a better understanding of global development dynamics. Ultimately, the project aims to improve the assessment of human development, leading to more effective policies and strategies that can enhance the quality of life across different countries and communities.

2.LITERATURE SURVEY

2.1 EXISTING PROBLEM

- **Income Inequality:**

Distribution Issues: The HDI uses average income levels and does not account for income inequality within a country. High income inequality can lead to a skewed understanding of a country's development.

- **Education and Health Metrics:**

Quality Over Quantity: The HDI uses quantitative measures (years of schooling and life expectancy) without considering the quality of education and healthcare services.

- **Neglect of Non-Income Indicators:**

Holistic Well-being: The HDI's emphasis on income as a significant component can overshadow other crucial non-income indicators of well-being, such as mental health, happiness, and social inclusion.

2.2 PROPOSED SOLLUTION

Our innovative proposed solution aim to create a more comprehensive, accurate, and actionable measure of human development.

1. Address Income Inequality:

- **Gini Coefficient or Palma Ratio:** Supplement GNI per capita with measures of income inequality to provide a more nuanced view of economic well-being.
- **Distribution-Adjusted Income:** Use income metrics adjusted for distribution to better reflect the economic reality of different population segments.

2. Enhance Education and Health Metrics:

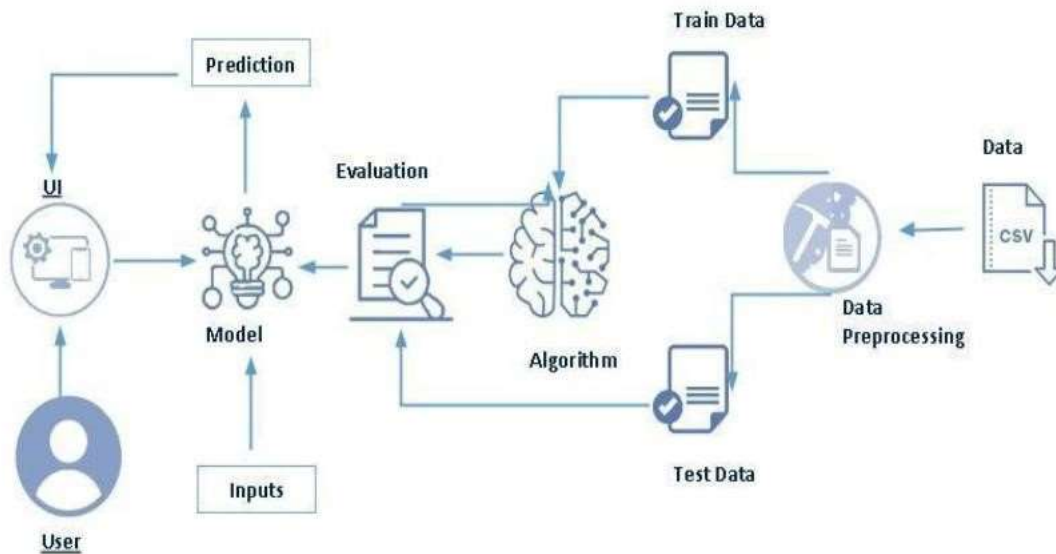
- **Quality of Education:** Include indicators like literacy rates, quality of education, student-teacher ratios, and learning outcomes.
- **Health Quality:** Consider healthcare accessibility, quality of healthcare services, and health outcomes beyond life expectancy, such as disease prevalence and mental health.

3. Incorporate Cultural and Societal Factors:

- **Cultural Sensitivity:** Include indicators that reflect cultural values and societal priorities, ensuring that the HDI is relevant to different cultural contexts.
- **Subjective Well-Being:** Integrate measures of happiness, life satisfaction, and subjective well-being to capture a holistic view of human development.
- By implementing these solutions, the HDI can be transformed into a more comprehensive, accurate, and actionable tool for measuring and promoting human development worldwide.

3.THEORITICAL ANALYSIS

3.1. BLOCK DIAGRAM



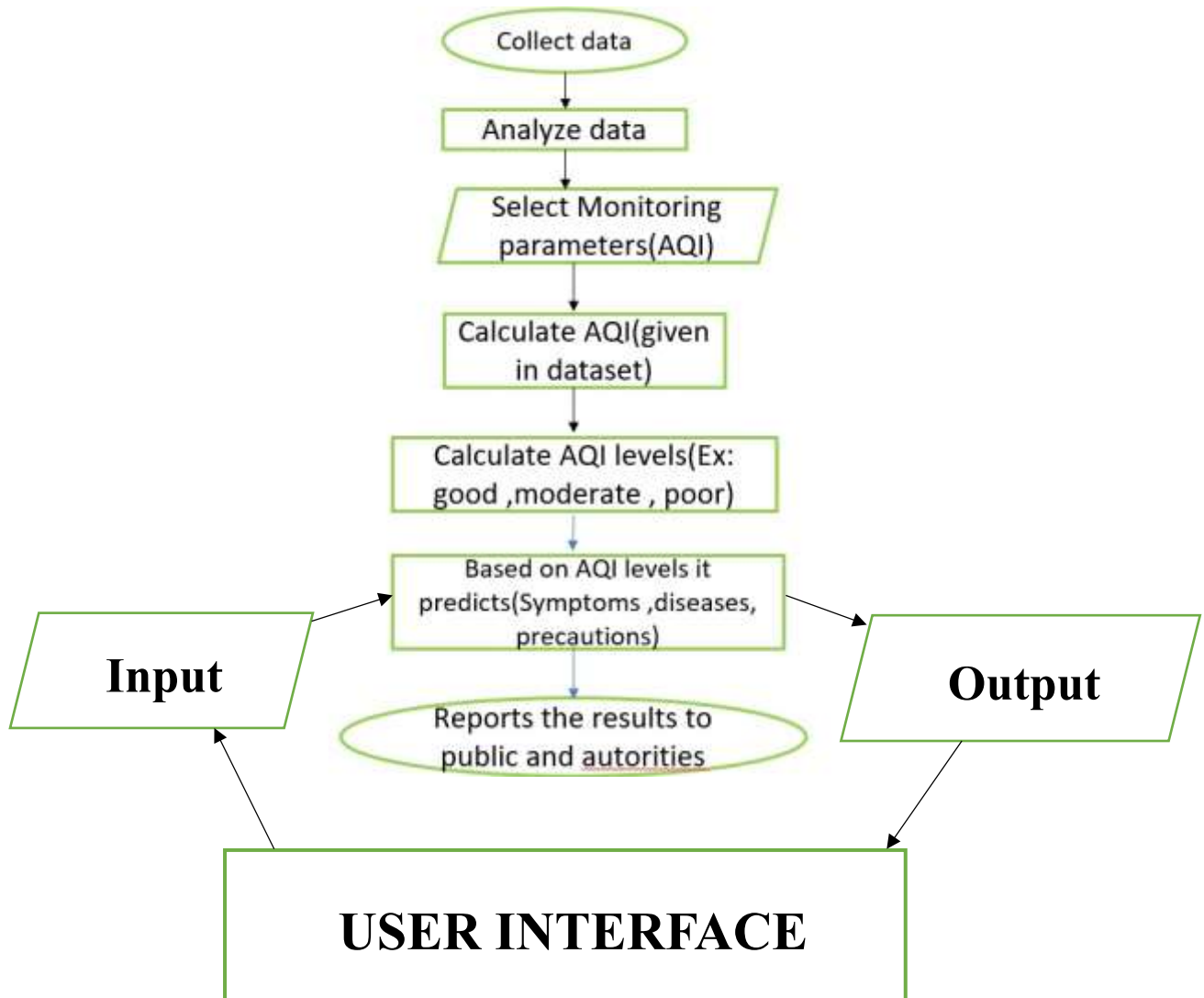
3.2. SOFTWARE DESIGNING

The following is the Software required to complete this project:

- **Google Colab:** Google Colab will serve as the development and execution environment for your predictive modeling, data preprocessing, and model training tasks. It provides a cloud-based Jupyter Notebook environment with access to Python libraries and hardware acceleration.
- **Dataset (CSV File):** The dataset in CSV format is essential for training and testing your predictive model. It should include historical air quality data, weather information, pollutant levels, and other relevant features.

- **Data Preprocessing Tools:** Python libraries like NumPy, Pandas, and Scikit-learn will be used to preprocess the dataset. This includes handling missing data, feature scaling, and data cleaning.
- **Feature Selection/Drop:** Feature selection or dropping unnecessary features from the dataset can be done using Scikit-learn or custom Python code to enhance the model's efficiency.
- **Model Training Tools:** Machine learning libraries such as Scikit-learn, TensorFlow, will be used to develop, train, and fine-tune the predictive model. Regression or classification models can be considered, depending on the nature of the HDI prediction task.
- **Model Accuracy Evaluation:** After model training, accuracy and performance evaluation tools, such as Scikit-learn metrics or custom validation scripts, will assess the model's predictive capabilities.
- **UI Based on Flask Environment:** Flask, a Python web framework, will be used to develop the user interface (UI) for the system. The Flask application will provide a user-friendly platform for users to input location data or view HDI predictions, health information, and recommended precautions.
- Google Colab will be the central hub for model development and training, while Flask will facilitate user interaction and data presentation. The dataset, along with data preprocessing, will ensure the quality of the training data, and feature selection will optimize the model. Finally, model accuracy evaluation will confirm the system's predictive capabilities, allowing users to rely on the HDI predictions.

5.FLOWCHART

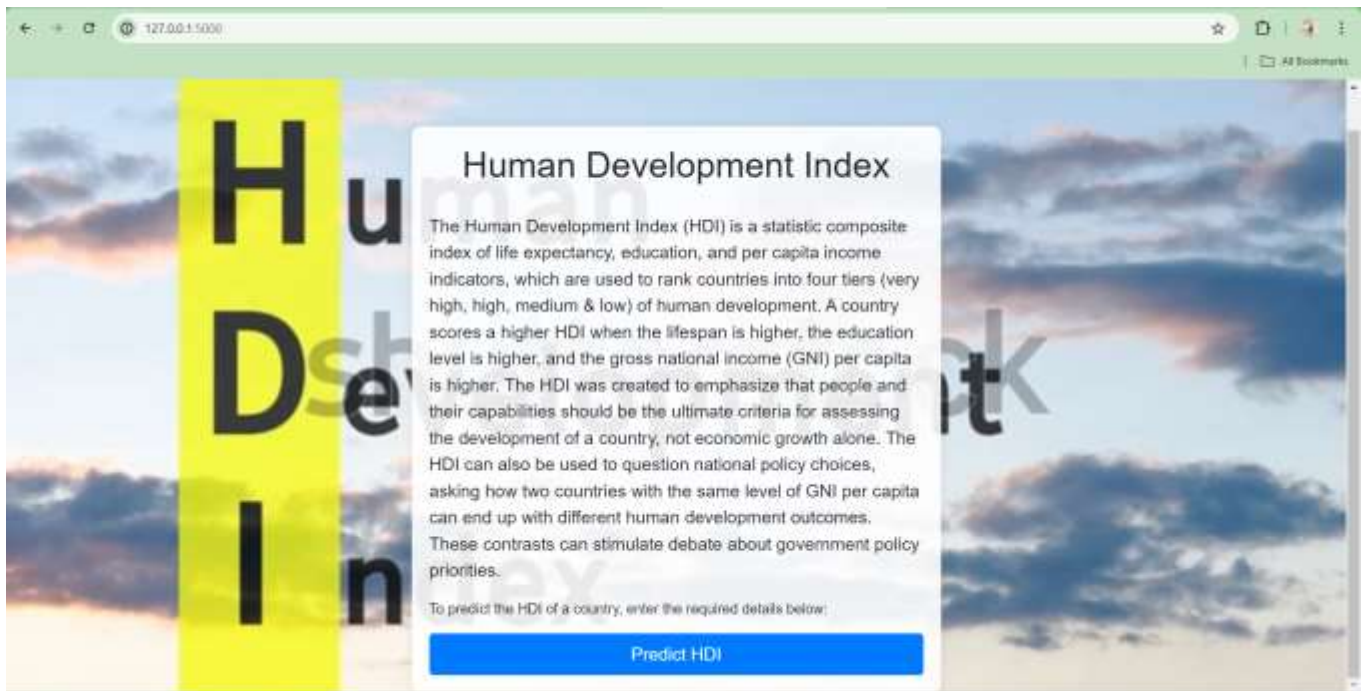


5.RESULT

Executing app.py :

```
Anaconda Prompt - python a
(base) C:\Users\pathi>cd C:\Users\pathi\OneDrive\Desktop\deekshith\flask
(base) C:\Users\pathi\OneDrive\Desktop\deekshith\flask>conda activate s1
(s1) C:\Users\pathi\OneDrive\Desktop\deekshith\flask>python app.py
* Serving Flask app 'app'
* Debug mode: on
WARNING: This is a development server. Do not use it in a production deployment. Use a production WSGI server instead.
* Running on http://127.0.0.1:5000
Press CTRL+C to quit
* Restarting with stat
* Debugger is active!
* Debugger PIN: 212-552-128
127.0.0.1 - - [15/Jul/2024 20:14:49] "GET / HTTP/1.1" 200 -
127.0.0.1 - - [15/Jul/2024 20:14:50] "GET /favicon.ico HTTP/1.1" 404 -
127.0.0.1 - - [15/Jul/2024 20:15:29] "GET /Prediction HTTP/1.1" 200 -
127.0.0.1 - - [15/Jul/2024 20:16:07] "POST /predict HTTP/1.1" 200 -
```

HOME PAGE



INDEX PAGE



Enter Details to Predict HDI

Life Expectancy:
82.17

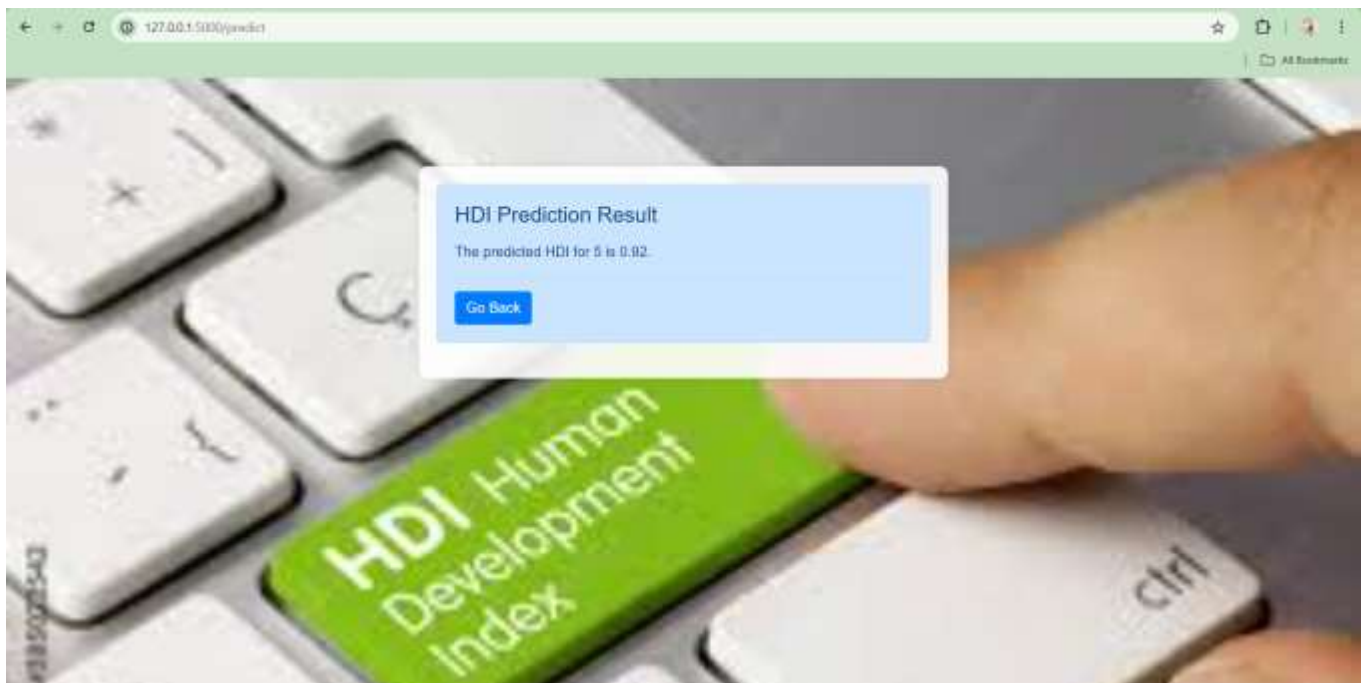
Mean Years of Schooling:
12.7

GNI per Capita:
60000.0

Country:
5

Predict HDI

PREDICTED PAGE



7.ADVANTAGES AND DISADVANTAGES

ADVANTAGES:

1. HDI measures three factors, rather than just income, giving a more rounded view of living standards than GDP.
2. Health, education and income are pivotal to good living standards
3. The data used is easy--to--obtain and reliable
4. Gives an idea about future living standards (i.e. education and life expectancy are indicators of the future situation)
5. The HDI is given as one number between 0 and 1, which allows for easy comparison between different countries.

DISADVANTAGES:

1. HDI gives no indication equality. Income and access to educational and health services might be high amongst a small group of people but low amongst others.
2. Years of schooling is unreliable if students are repeating years due to lack of progress. Also, going to school is worthless if the education is of a poor quality.
3. The calculation excludes outliers from the data, which is unfair because countries should be rewarded with higher HDIs for good performance in one particular field.
4. Adult literacy places too much emphasis on reading and writing-- in this modern age computer literacy is arguably more important.

8.APPLICATIONS

1. **Policy Making:** Governments use the HDI to identify areas that needed improvement in health, education and income to enhance overall human development.
2. **Research and Analysis:** Researchers and academics use the HDI to study trends in human development over time and across different regions, providing valuable insights for policy recommendations.
3. **Global Ranking:** The HDI is used to rank countries based on their development levels, allowing for comparison and highlighting disparities that need attention.
4. **Sustainable Development Goals:** The HDI is used to track progress towards achieving the United Nation's sustainable development goals, which aim to improve the quality of life for people around the world.

9.CONCLUSION

- In conclusion, the Human Development Index (HDI) serves as a pivotal tool in the global landscape of development measurement, offering a multifaceted view of human well-being that transcends mere economic metrics. By incorporating health, education, and income indicators, the HDI provides a comprehensive snapshot of development that is easy to understand and widely accepted. It guides policymakers, informs international comparisons, and raises public awareness about the crucial dimensions of human development.
- while the HDI is an invaluable tool for assessing and promoting human development, ongoing efforts to expand and improve its methodology are essential. These improvements will ensure that the HDI continues to serve as a reliable and comprehensive measure, driving policies and actions that enhance the quality of life and well-being for people worldwide.

10.FUTURE SCOPE

Future Scope of the A COMPREHENSIVE MEASURE OF WELL-BEING: A HUMAN DEVELOPMENT INDEX are:

1. **Environmental Sustainability:** Integrate indicators related to environmental health and sustainability, such as carbon emissions, air and water quality, and resource use.
2. **Multidimensional Poverty Index (MPI):** Complement the HDI with MPI to capture the broader spectrum of poverty and inequality..
3. **Income Distribution Metrics:** Incorporate measures like the Gini coefficient or Palma ratio to reflect income inequality.
4. **Alignment with SDGs:** Ensure the HDI is aligned with the United Nations Sustainable Development Goals, providing a comprehensive framework for measuring progress toward these goals.

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12.APPENDIX

Model building :

1)Dataset

2)Google Colab and VS code Application Building

- HTML file (home.html, indexnew.html, resultsnew.html, error.html)
- Models in pickle format

SOURCE CODE:

Home.html :

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <title>HDI Predictor</title>
  <link
href="https://stackpath.bootstrapcdn.com/bootstrap/4.5.2/css/bootstrap.min.css"
rel="stylesheet">
  <style>
    body {
      font-family: Arial, sans-serif;
      background-image: url('https://www.shutterstock.com/image-
illustration/acronym-hdi-human-development-index-600w-339214301.jpg'); /* Replace with
your background image URL */
      background-size: cover;
      background-position: center;
      background-repeat: no-repeat;
      background-attachment: fixed;
    }

    .container {
      max-width: 600px;
```

```

        margin-top: 100px;
        background-color: rgba(255, 255, 255, 0.9);
        padding: 20px;
        border-radius: 10px;
        box-shadow: 0px 0px 10px rgba(0, 0, 0, 0.1);
    }

    h1 {
        text-align: center;
        margin-bottom: 30px;
    }

    .btn-primary {
        background-color: #007bff;
        border-color: #007bff;
    }

    .btn-primary:hover {
        background-color: #0069d9;
        border-color: #0062cc;
    }
</style>
</head>
<body>
    <div class="container">
        <h1 class="text-center">Human Development Index</h1>
        <p class="lead">The Human Development Index (HDI) is a statistic composite
index of life expectancy, education, and per capita income indicators, which are used
to rank countries into four tiers (very high, high, medium & low) of human development.
A country scores a higher HDI when the lifespan is higher, the education level is
higher, and the gross national income (GNI) per capita is higher. The HDI was created
to emphasize that people and their capabilities should be the ultimate criteria for
assessing the development of a country, not economic growth alone. The HDI can also be
used to question national policy choices, asking how two countries with the same level
of GNI per capita can end up with different human development outcomes. These contrasts
can stimulate debate about government policy priorities.</p>
        <p>To predict the HDI of a country, enter the required details below:</p>
        <a href="/Prediction" class="btn btn-primary btn-lg btn-block">Predict HDI</a>
    </div>

    <script src="https://code.jquery.com/jquery-3.5.1.slim.min.js"></script>
    <script
src="https://cdn.jsdelivr.net/npm/@popperjs/core@2.9.2/dist/umd/popper.min.js"></script
>

```

```

    <script
src="https://stackpath.bootstrapcdn.com/bootstrap/4.5.2/js/bootstrap.min.js"></script>
</body>
</html>

```

Indexnew.html :

```

<!DOCTYPE html>
<html lang="en">
<head>
    <meta charset="UTF-8">
    <meta name="viewport" content="width=device-width, initial-scale=1.0">
    <title>Predict HDI</title>
    <link
href="https://stackpath.bootstrapcdn.com/bootstrap/4.5.2/css/bootstrap.min.css"
rel="stylesheet">
    <style>
        body {
            font-family: Arial, sans-serif;
            background-image: url('https://encrypted-
tbn0.gstatic.com/images?q=tbn:ANd9GcQovEiSw8rQFuFN29sWsVJwnqY7ijwtzFenRw&s'); /*
Replace with your background image URL */
            background-size: cover;
            background-position: center;
            background-repeat: no-repeat;
            background-attachment: fixed;
        }

        .container {
            max-width: 600px;
            margin-top: 100px;
            background-color: rgba(255, 255, 255, 0.9);
            padding: 20px;
            border-radius: 10px;
            box-shadow: 0px 0px 10px rgba(0, 0, 0, 0.1);
        }

        h1 {
            text-align: center;
            margin-bottom: 30px;
        }

        .btn-primary {

```

```

        background-color: #007bff;
        border-color: #007bff;
    }

    .btn-primary:hover {
        background-color: #0069d9;
        border-color: #0062cc;
    }
</style>
</head>
<body>
    <div class="container">
        <h1 class="text-center">Enter Details to Predict HDI</h1>
        <form action="/predict" method="post">
            <div class="form-group">
                <label for="lifeExpectancy">Life Expectancy:</label>
                <input type="text" class="form-control" id="lifeExpectancy"
name="lifeExpectancy" required>
            </div>
            <div class="form-group">
                <label for="meanYearsOfSchooling">Mean Years of Schooling:</label>
                <input type="text" class="form-control" id="meanYearsOfSchooling"
name="meanYearsOfSchooling" required>
            </div>
            <div class="form-group">
                <label for="gniPerCapita">GNI per Capita:</label>
                <input type="text" class="form-control" id="gniPerCapita"
name="gniPerCapita" required>
            </div>
            <div class="form-group">
                <label for="Country">Country:</label>
                <input type="text" class="form-control" id="Country" name="Country"
required>
            </div>
            <button type="submit" class="btn btn-primary btn-block">Predict
HDI</button>
        </form>
    </div>

    <script src="https://code.jquery.com/jquery-3.5.1.slim.min.js"></script>
    <script
src="https://cdn.jsdelivr.net/npm/@popperjs/core@2.9.2/dist/umd/popper.min.js"></script>
    <script
src="https://stackpath.bootstrapcdn.com/bootstrap/4.5.2/js/bootstrap.min.js"></script>

```



```

    h4 {
        margin-bottom: 15px;
    }

    .btn-primary {
        background-color: #007bff;
        border-color: #007bff;
    }

    .btn-primary:hover {
        background-color: #0069d9;
        border-color: #0062cc;
    }
</style>
</head>
<body>
    <div class="container">
        <div class="alert alert-info" role="alert">
            <h4 class="alert-heading">HDI Prediction Result</h4>
            <p>{{ prediction_text }}</p>
            <hr>
            <a href="/home" class="btn btn-primary">Go Back</a>
        </div>
    </div>

    <script src="https://code.jquery.com/jquery-3.5.1.slim.min.js"></script>
    <script
src="https://cdn.jsdelivr.net/npm/@popperjs/core@2.9.2/dist/umd/popper.min.js"></script>
    <script
src="https://stackpath.bootstrapcdn.com/bootstrap/4.5.2/js/bootstrap.min.js"></script>
</body>
</html>

```

Error.html :

```

<!DOCTYPE html>
<html lang="en">
<head>
    <meta charset="UTF-8">
    <meta name="viewport" content="width=device-width, initial-scale=1.0">
    <title>Error</title>
    <style>

```

```

    body, html {
        height: 100%;
        margin: 0;
        font-family: Arial, sans-serif;
        text-align: center;
        background-image: url('background-image.jpg'); /* Replace with your
background image */
        background-position: center;
        background-repeat: no-repeat;
        background-size: cover;
        display: flex;
        justify-content: center;
        align-items: center;
        color: white;
    }
    .container {
        background-color: rgba(0, 0, 0, 0.7); /* Dark overlay */
        padding: 40px;
        border-radius: 10px;
        width: 50%;
    }
    h1 {
        font-size: 2.5em;
        color: red;
        margin-bottom: 20px;
    }
    p {
        font-size: 1.2em;
        color: yellowgreen;
    }
</style>
</head>
<body>
    <div class="container">
        <h1>Error</h1>
        <p>{{ error_text }}</p> <!-- Dynamic error message -->
    </div>
</body>
</html>

```

App.py :

```

from flask import Flask, render_template, request, redirect, url_for

```

```

import pickle

app = Flask(__name__)

# Load the model
with open('HDI.pkl', 'rb') as f:
    model = pickle.load(f)

@app.route('/')
@app.route('/home')
def home():
    return render_template('home.html')

@app.route('/Prediction')
def prediction():
    return render_template('indexnew.html')

@app.route('/predict', methods=['POST'])
def predict():
    try:
        # Retrieve form data
        life_expectancy = float(request.form['lifeExpectancy'])
        mean_years_of_schooling = float(request.form['meanYearsOfSchooling'])
        gni_per_capita = float(request.form['gniPerCapita'])
        country = request.form['Country']

        # Assuming the model requires these inputs in a specific order
        features = [life_expectancy, mean_years_of_schooling, gni_per_capita, country]

        # Make prediction
        prediction = model.predict([features])[0]
        prediction_text = f'The predicted HDI for {country} is {prediction:.2f}.'

        return render_template('resultsnew.html', prediction_text=prediction_text)
    except Exception as e:
        error_text = str(e)
        return render_template('error.html', error_text=error_text)

if __name__ == '__main__':
    app.run(debug=True)

```


CODE SNIPPETS

MODEL BUILDING :

HDUipynb

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Code Text

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```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from plotly import offline
```

```
data = pd.read_csv("../content/HDI.csv")
```

```
data.head()
```

Unnamed: 0	Id	Country	HDI Rank	HDI	Life expectancy	Mean years of schooling	Gross national income (GNI) per capita	GNI per capita rank	Change in HDI rank 2010-2015	Coefficient of human inequality	Inequality in life expectancy (%) 2010-2015	Inequality-adjusted life expectancy index	Inequality in education (%)	Inequality-adjusted education index	Inequality in income (%)	Inequality-adjusted income index	Income inequality (Quintile ratio) 2010-2015	Income inequality (Palme ratio) 2010-2015	
0	0	1	Norway	1.0	0.949	81.7	12.7	67814.0	5.0	0.0	5.4	3.3	0.010	2.4	0.004	10.4	0.002	3.8	0.9
1	1	2	Australia	2.0	0.938	82.5	13.2	42822.0	18.0	1.0	4.0	4.3	0.021	1.9	0.021	17.7	0.753	6.0	1.4
2	2	3	Switzerland	2.0	0.909	83.1	13.4	96364.0	7.0	0.0	8.4	3.8	0.034	5.7	0.040	15.7	0.006	4.9	5.2
3	3	4	Germany	4.0	0.828	81.1	13.2	45000.0	13.0	0.0	7.0	3.7	0.005	2.8	0.001	14.8	0.787	4.6	1.1
4	4	5	Denmark	5.0	0.925	80.4	12.7	44510.0	13.0	2.0	7.0	3.8	0.004	3.0	0.006	14.3	0.789	4.0	1.0

```
$ rows = 82 columns
```

```
data.tail()
```

```
data.tail()
```

190	190	191	Morocco	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
191	191	192	Nauru	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
192	192	193	San Marino	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
193	193	194	Somalia	NaN	NaN	55.7	NaN	294.0	NaN	NaN	NaN	42.1	0.318	43.5	NaN	NaN	NaN	NaN	NaN
194	194	195	Tuvalu	NaN	NaN	NaN	NaN	5595.0	NaN	NaN	NaN	NaN	10.5	NaN	NaN	NaN	7.7	1.9	NaN

```
$ rows = 82 columns
```

The screenshot shows a Jupyter Notebook with a single cell containing the following code:

```
data.info()

class 'pandas.core.frame.DataFrame'
Int64Index: 190 entries, 0 to 190
Data columns (total 22 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   Unnamed: 0                            190 non-null    int64
 1   Id                                     190 non-null    int64
 2   Country                               190 non-null    object
 3   HDI Rank                              188 non-null    float64
 4   HDI                                    188 non-null    float64
 5   Life expectancy                       190 non-null    float64
 6   Mean years of schooling                188 non-null    float64
 7   Gross national income (GNI) per capita 190 non-null    float64
 8   GNI per capita rank minus HDI rank     188 non-null    float64
 9   Change in HDI rank 2010-2015          186 non-null    float64
10  Average annual HDI growth 1990-2008   144 non-null    float64
11  Average annual HDI growth 2000-2008   168 non-null    float64
12  Average annual HDI growth 2010-2015   188 non-null    float64
13  Average annual HDI growth 1990-2015   142 non-null    float64
14  Gender Development Index value         160 non-null    float64
15  Gender Development Index Group         160 non-null    float64
16  Human Development Index (HDI) Female   168 non-null    float64
17  Human Development Index (HDI) Male     168 non-null    float64
18  Life expectancy at birth Female        185 non-null    float64
19  Life expectancy at birth Male          185 non-null    float64
20  Mean years of schooling Female         160 non-null    float64
21  Mean years of schooling Male           160 non-null    float64
22  Estimated gross national income per capita Female 178 non-null    float64
23  Estimated gross national income per capita Male 178 non-null    float64
24  Share of seats in parliament (% held by women) 192 non-null    float64
25  Population with at least some secondary education (% (2000-2015)) Female 164 non-null    float64
26  Population with at least some secondary education (% (2000-2015)) Male 164 non-null    float64
27  Labor force participation rate (% ages 15 and older) Female 180 non-null    float64
28  Total Population (millions) 2019       190 non-null    float64
29  Total Population (millions) 2030       190 non-null    float64
```

HDUipymb

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```
data=data.nlargest(10,"HDI")
```

data1

Unnamed: 0	Id	Country	HDI Rank	HDI	Life expectancy	Mean years of schooling	Gross national income (GNI) per capita	GNI per capita rank minus HDI rank	Change in HDI rank 2010-2015	Coefficient of human inequality	Inequality in life expectancy (%) 2010-2015	Inequality-adjusted life expectancy index	Inequality in education (%)	Inequality-adjusted education index	Inequality in income (%)	Inequality-adjusted income index	Income inequality (Quintile ratio) 2010-2015	Income inequality (Palme ratio) 2010-2015
0	0	1	Norway	1.0	0.949	81.7	12.7	67614.0	-5.0	0.0	5.4	3.3	0.018	2.4	0.004	10.4	0.002	3.8
1	1	2	Australia	2.0	0.939	82.5	13.2	42822.0	10.0	1.0	8.0	4.3	0.021	1.9	0.021	17.7	0.753	0.0
2	2	3	Switzerland	2.0	0.939	83.1	13.4	56364.0	7.0	0.0	8.4	3.8	0.004	5.7	0.040	15.7	0.006	4.0
3	3	4	Germany	4.0	0.920	81.1	13.3	45000.0	13.0	0.0	7.0	3.7	0.005	2.6	0.001	14.8	0.707	4.8
4	4	5	Denmark	5.0	0.925	80.4	12.7	44519.0	13.0	2.0	7.0	3.8	0.004	3.0	0.006	14.3	0.789	4.5
5	5	6	Singapore	5.0	0.925	83.2	11.8	78162.0	-3.6	0.0	NaN	3.0	0.043	NaN	NaN	NaN	NaN	NaN
6	6	7	Netherlands	7.0	0.924	81.7	11.9	46326.0	8.0	-2.0	6.8	3.7	0.014	4.2	0.058	12.4	0.812	4.2
7	7	8	Iceland	8.0	0.923	81.1	12.5	43708.0	11.0	1.0	7.7	3.7	0.005	3.0	0.003	16.3	0.760	5.3
8	8	9	Ireland	8.0	0.921	82.7	12.2	37065.0	20.0	7.0	5.7	2.8	0.007	2.5	0.004	11.7	0.709	4.0
9	9	10	Canada	10.0	0.920	82.2	13.1	42562.0	12.0	1.0	8.7	4.7	0.012	3.9	0.056	17.4	0.755	5.8
10	10	11	United States	10.0	0.920	79.2	13.2	53245.0	1.0	-3.0	12.0	6.1	0.058	5.6	0.050	27.0	0.602	9.1
11	11	12	Hong Kong (China (HKSAR))	12.0	0.917	84.2	11.8	54266.0	-2.0	3.0	NaN	2.8	0.053	NaN	NaN	NaN	NaN	NaN

HDUipymb

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+ Code + Text

```
from sklearn.preprocessing import LabelEncoder
```

lr=LabelEncoder()

```
for Country in data1.select_dtypes(include=['object']).columns:
    data1[Country]=lr.fit_transform(data1[Country])
```

```
import pandas as pd
from sklearn.preprocessing import OneHotEncoder

# Identify categorical columns
categorical_columns = data1.select_dtypes(include=['object']).columns

# Apply One-hot encoding to categorical columns
data_encoded = pd.get_dummies(data1, columns=categorical_columns, drop_first=True)
```

```
# Display the first few rows of the encoded dataset
data_encoded.head()
```

HDUipymb

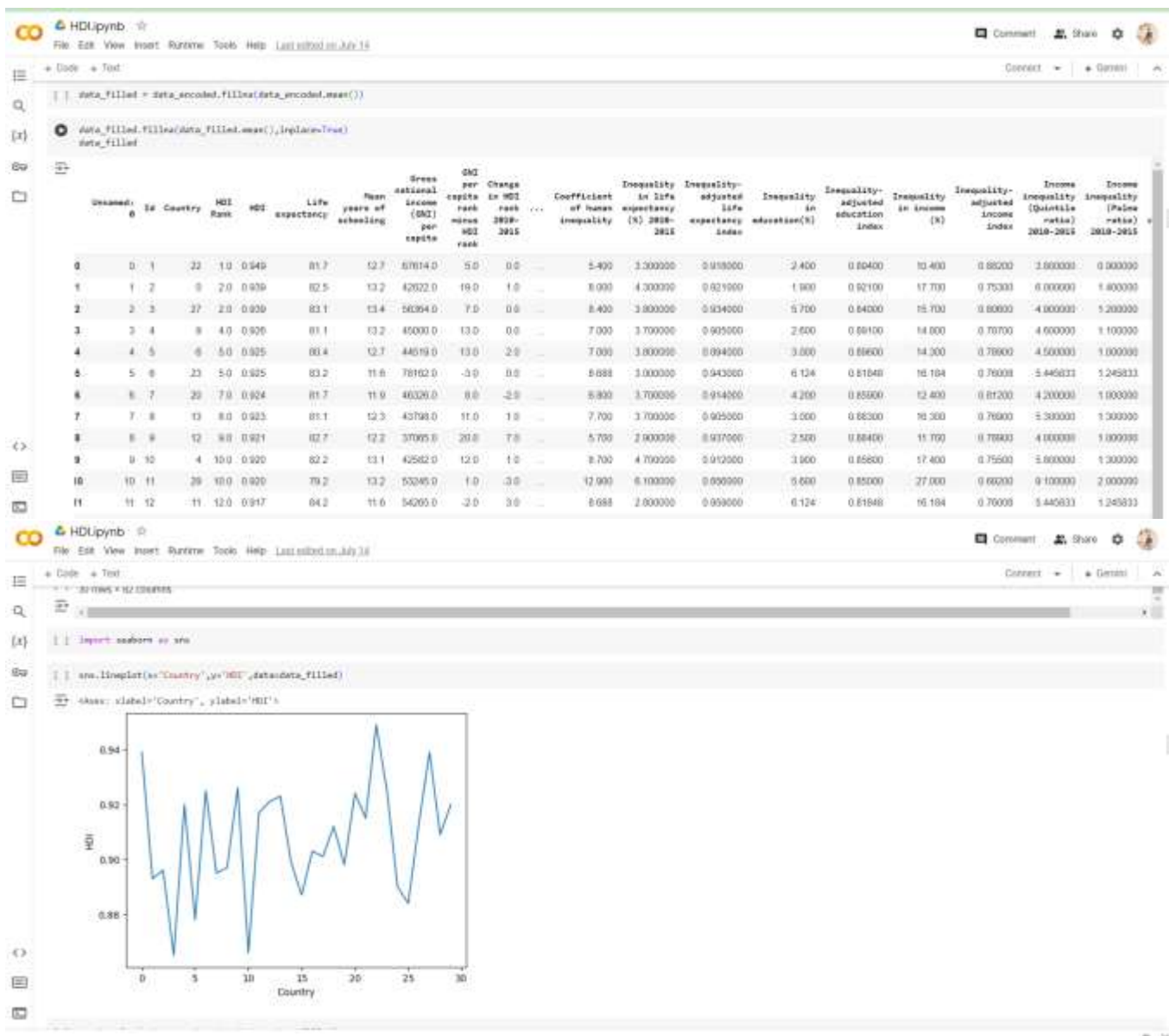
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+ Code + Text

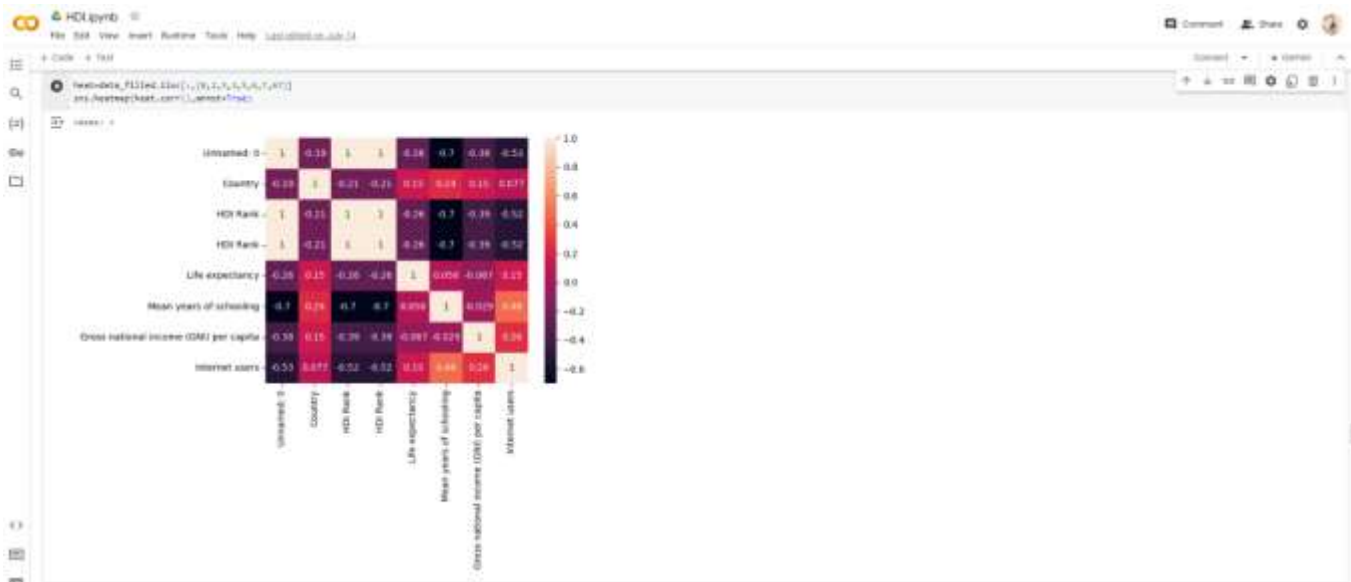
```
# Display the first few rows of the encoded dataset
data_encoded.head()
```

Unnamed: 0	Id	Country	HDI Rank	HDI	Life expectancy	Mean years of schooling	Gross national income (GNI) per capita	GNI per capita rank minus HDI rank	Change in HDI rank 2010-2015	Coefficient of human inequality	Inequality in life expectancy (%) 2010-2015	Inequality-adjusted life expectancy index	Inequality in education (%)	Inequality-adjusted education index	Inequality in income (%)	Inequality-adjusted income index	Income inequality (Quintile ratio) 2010-2015	Income inequality (Palme ratio) 2010-2015
0	0	1	22	1.0	0.949	81.7	12.7	67614.0	3.0	0.0	5.4	3.3	0.018	2.4	0.004	10.4	0.002	3.8
1	1	2	0	2.0	0.939	82.5	13.2	42822.0	19.0	1.0	8.0	4.3	0.021	1.9	0.021	17.7	0.753	0.0
2	2	3	27	2.0	0.939	83.1	13.4	56364.0	7.0	0.0	8.4	3.8	0.004	5.7	0.040	15.7	0.006	4.0
3	3	4	9	4.0	0.920	81.1	13.2	45000.0	13.0	0.0	7.0	3.7	0.005	2.6	0.001	14.8	0.707	4.8
4	4	5	8	5.0	0.925	80.4	12.7	44519.0	13.0	2.0	7.0	3.8	0.004	3.0	0.006	14.3	0.789	4.5

5 rows x 19 columns







```

import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score

# Loading 'data' in your original dataframe
x = data.iloc[:, [2, 3, 4, 5, 6, 7]]
y = data.iloc[:, 1].values

# Splitting the data into training and testing sets
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=0)

# Create the Linear Regression model
regressor = LinearRegression()

# Training the model with the training set
regressor.fit(x_train, y_train)

# Predicting the test set results
y_pred = regressor.predict(x_test)

# Calculating the R-squared value
r2 = r2_score(y_test, y_pred)
print(r2)

```

```

# Predicting the test set results
y_pred = regressor.predict(x_test)

# Calculating the R-squared value
r2 = r2_score(y_test, y_pred)
print(r2)

```


PAIRPLOT

HEAT MAP

DESCRIPTIVE ANALYSIS