

Project Overview

This project aims to analyze the global state of 'happiness' in the world, from 2006 to 2020, from people in over 150 countries, using the United Nations Happiness Index dataset.



Data Story

Show the correlation between measurements of well-being and create a model to identify which one is essential to discovering the 'key to happiness'.



Data Source

The raw dataset files were obtained from the Kaggle website as csv files.



Target Audience

Students and research professionals.

How It Was Done

01

ETL/Web Design

02

ML Model/ Database Structure

03

Python "Flask" App

04

Heroku Deployment

Data Processing

After establishing our data sources and dependencies, the ETL was implemented.

- null values/duplicates were removed
- columns renamed accordingly
- Groupby/sort/filter data accordingly

RAW DATA

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1949 entries, 0 to 1948
Data columns (total 11 columns):
    Column
                                       Non-Null Count Dtype
    Country name
                                       1949 non-null
                                                       object
    Life Ladder
                                                       float64
    Log GDP per capita
    Social support
    Healthy life expectancy at birth 1894 non-null
                                                       float64
    Freedom to make life choices
                                                       float64
                                       1917 non-null
                                       1860 non-null
                                                       float64
                                                       float64
    Perceptions of corruption
                                       1839 non-null
    Positive affect
                                                       float64
                                       1927 non-null
    Negative affect
                                       1933 non-null
                                                       float64
dtypes: float64(9), int64(1), object(1)
memory usage: 167.6+ KB
```

TRANSFORMED DATA

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1708 entries, 0 to 1948
Data columns (total 11 columns):
    Column
                                       Non-Null Count Dtype
                                       1708 non-null
                                                      object
    Country name
                                                      int64
    Life Ladder
                                                      float64
     Log GDP per capita
                                       1708 non-null
                                                       float64
     Social support
                                      1708 non-null
                                                      float64
    Healthy life expectancy at birth 1708 non-null
                                                      float64
     Freedom to make life choices
                                       1708 non-null
                                                       float64
     Generosity
                                       1708 non-null
                                                      float64
    Perceptions of corruption
                                                      float64
    Positive affect
                                      1708 non-null
                                                      float64
10 Negative affect
                                       1708 non-null
                                                      float64
dtypes: float64(9), int64(1), object(1)
memory usage: 160.1+ KB
```

ML Model

Scikit-learn library was used to build our data model.

- Regression Trees (rt) was chosen to work with our numeric target variables.
- train_test_split was used to create training and testing data (80% train, 20% test).
- Fitted the model to the training data.
- Tested the model to the unseen test data.
- The model was used to make predictions for 2021 data.

VARIABLES

```
X = happinessByYears_df[['GDP_per_capita','Life_expectancy']]
y = happinessByYears_df['Life_ladder'].values.reshape(-1, 1)
print(X.shape, y.shape)
(1708, 2) (1708, 1)
```

MODEL

```
rt = DecisionTreeRegressor(criterion = 'mse', max_depth=2)
rt
```

PREDICTIONS

```
Prediction Actual

0 7.159857 7.842

1 7.159857 7.620

2 7.159857 7.554

4 7.159857 7.554

4 7.159857 7.464

...

144 4.357920 3.512

145 4.357920 3.467

146 5.368271 3.415

147 4.357920 3.145

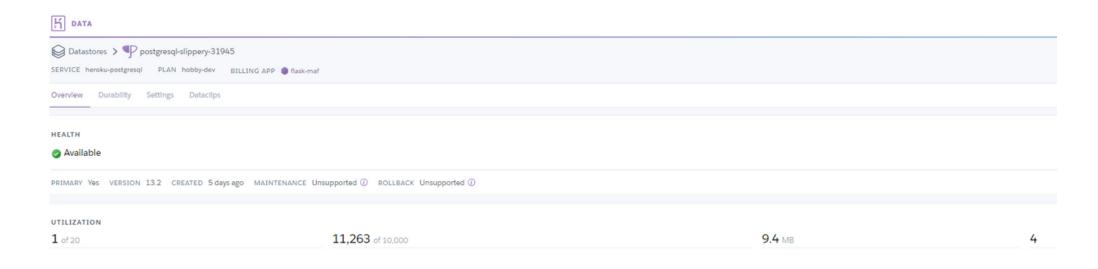
148 4.357920 2.523

149 rows × 2 columns
```

Database

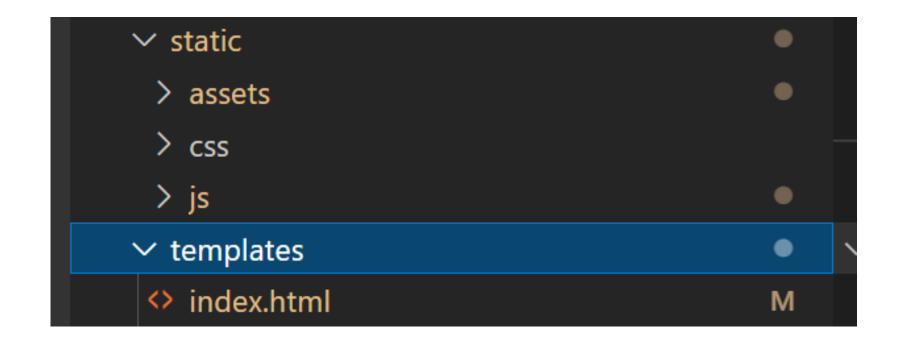
The transformed main dataframes and predictions dataframe were successfully loaded to the PostgreSQL database using sqlalchemy

```
happinessByYears_df.to_sql(name='happinessoveryears', con=engine, if_exists='append', index=False)
statistics.to_sql(name='statistics', con=engine, if_exists='append', index=False)
gbMainData.to_sql(name='gbmaindata', con=engine, if_exists='append', index=False)
bottom.to_sql(name='bottom', con=engine, if_exists='append', index=False)
top.to_sql(name='top', con=engine, if_exists='append', index=False)
```



Web Design

JavaScript, HTML, CSS from a Bootstrap template were used for this section. Changes were made accordingly.



Flask

Connection to the database was done via flask API Calls. API routes for each table were created to get the data in json format for the visualizations.

```
ect4 > 🕏 main.py > ...
        gbMainData_df_json = json.dumps(parsed, skipkeys = True, allow_nan = True, indent = 6)
        return gbMainData df json
  @app.route("/api/bottom")
  def bottom():
      result = bottom_df.to_json(orient="records")
      parsed = json.loads(result)
      bottom_df_json = json.dumps(parsed, skipkeys = True, allow_nan = True, indent = 6)
      return bottom_df_json
  @app.route("/api/top")
  def top():
      result = top_df.to_json(orient="records")
      parsed = json.loads(result)
      top_df_json = json.dumps(parsed, skipkeys = True, allow_nan = True, indent = 6)
      return top_df_json
  @app.route("/api/happiness2021")
  def happiness2021():
      result = happiness2021_df.to_json(orient="records")
      parsed = json.loads(result)
      happiness2021_df_json = json.dumps(parsed, skipkeys = True, allow_nan = True, indent = 6)
      return happiness2021_df_json
```

Visualizations

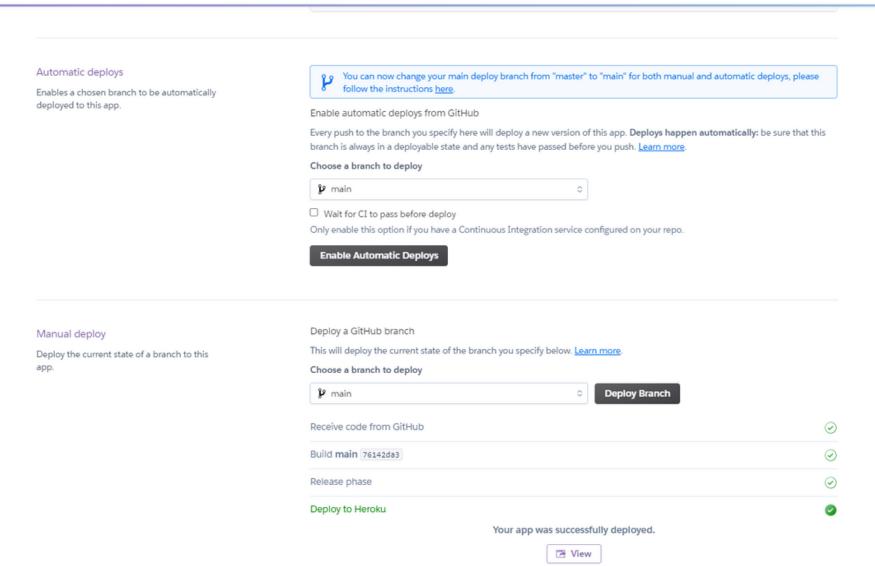
async function was used in our JavaScript visualization file to connect to the database via API call.

For Loop was used to get each individual variables from the json objects for the traces.

D3 and Plotly were used for the charts/maps

```
const api url main = '/api/main'
     async function getData main(){
     const response_main = await fetch(api_url_main)
     const data_main = await response_main.json();
     console.log(data main)
     let x1 = []
     let vl = []
     let Country = []
     let Corruption = []
     let Freedom =[]
     let GDP_per_capita =[]
     let Generosity =[]
     let Life_expectancy =[]
     let Social_support = []
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     for (var i=0; i< data_main.length; i++){</pre>
     xl.push(data_main[i]['year'])
     yl.push(data_main[i]['Life_ladder'])
     Corruption.push(data_main[i]['Corruption'])
     Freedom.push(data_main[i]['Freedom'])
     GDP_per_capita.push(data_main[i]['GDP_per_capita'])
     Generosity.push(data_main[i]['Generosity'])
     Country.push(data_main[i]['Country'])
     Social_support.push(data_main[i]['Social_support'])
     Life_expectancy.push(data_main[i]['Life_expectancy'])}
29
     // add choropleth map (avg Life_ladder/ years & countries)
     var data = [{
```

Heroku Deployment



Questions?