

+ Yargi, Emily, Anna and Archit +

Group 6



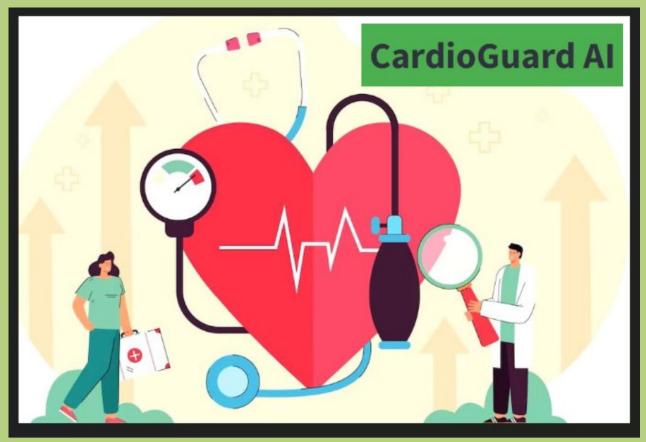
### CardioGuard Al: Predicting Heart Disease

Risk & Providing Al Lifestyle Advice

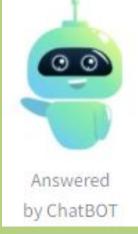


Life-Style Advices Powered by OpenALAPI











# Target Audience

#### Healthcare Professionals

- **Early Detection Tool:** Assists in identifying potential heart disease risk.
- Informed Interventions: Guides on when to consider further medical evaluation.

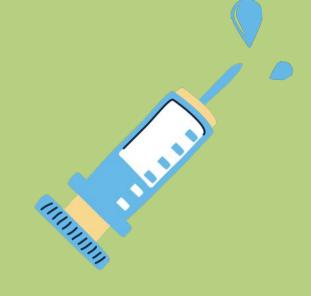
### Heart health conscious people

- Educational Insight: Provides preliminary assessment of heart disease risk factors.
- Guided Advice: Offers recommendations for lifestyle adjustments and seeking professional care.





### Data Sourcing



### **Data Source**



- Data Repository Name:
  - Kaggle
- Dataset Name: Heart
   Failure Prediction Dataset
- URL: <u>Heart Failure</u>
   <u>Prediction Dataset</u>

### **Key Dataset Features**



- **Records:** 918
- Features: 12
- Target: Heart Disease



## Data Cleaning, Modelling and Preprocessing



```
# heart data = pd.read sql('SELECT * FROM heart data', engine)
heart data = pd.read sql('SELECT * FROM heart data', engine)
# Display sample data
heart data.head()
     sex chestpaintype restingbp cholesterol fastingbs restingecg maxhr exerciseangina oldpeak st_slope heartdisease
                   ATA
                             140
                                                                    172
                                                                                                    Up
                                                                                                                  0
                  NAP
                             160
                                                                    156
                                                                                           1.0
                                                          Normal
                                                                                                                  0
                  ATA
                             130
                                        283
                                                             ST
                                                                     98
                                                                                           0.0
                                                                                                    Up
 48
                  ASY
                             138
                                        214
                                                          Normal
                                                                    108
                                                                                           1.5
                                                                                                    Flat
54
      М
                  NAP
                             150
                                        195
                                                                    122
                                                                                           0.0
                                                                                                                  0
                                                          Normal
                                                                                                    Up
```

```
# Looking at heartdisease value counts for binning
application_counts = heart_data.heartdisease.value_counts()
application_counts

heartdisease
1 508
0 410
Name: count, dtype: int64
```

We have a good split between instances that indicate heart disease and instances that indicate no heart disease.

Our dataset was very clean, limited cleaning was needed. However categorical features were mapped as a preprocessing step.

```
# Mapping for categorical features
sex_mapping = {'M': 1, 'F': 0}
cp_mapping = {'NAP': 0, 'ATA': 1, 'ASY': 2, 'TA': 3}
bs_mapping = {'Normal': 0, 'ST': 1}
ecg_mapping = {'Normal': 0, 'ST': 1}
angina_mapping = {'N': 0, 'Y': 1}
slope_mapping = {'Up': 0, 'Flat': 1, 'Down': 2}

# Convert categorical columns to numeric in the original dataset
heart_data['sex'] = heart_data['sex'].map(sex_mapping)
heart_data['chestpaintype'] = heart_data['chestpaintype'].map(cp_mapping)
heart_data['fastingbs'] = heart_data['fastingbs'].map(bs_mapping)
heart_data['restingecg'] = heart_data['restingecg'].map(ecg_mapping)
heart_data['exerciseangina'] = heart_data['exerciseangina'].map(angina_mapping)
heart_data['st_slope'] = heart_data['st_slope'].map(slope_mapping)
```



### Use of Elephant SQL - Cloud based



### Elephant SQL features

- Cloud SQL software
- Free plan available
- Administers tasks of PostgreSQL,
   including installation, upgrades and
   backup handling
- Allows us to have user interaction and cloud connection

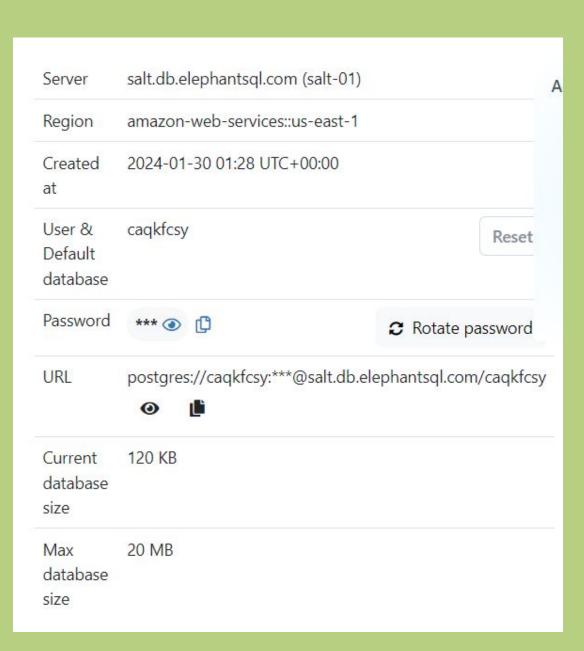
Use in code

```
# ElephantSQL database URI
database_uri = 'postgresql://caqkfcsy:b41wIKaMBjOtBNDe1Y6EL_Xz-uoGqvuU@salt.db.elephantsql.com/caqkfcsy'

# Create the database engine
engine = create_engine(database_uri)

# Load the dataset from PostgreSQL
heart_data = pd.read_sql('SELECT * FROM heart_data', engine)

# Rename columns based on PostgreSQL column names
```



### Machine learning models

1st

### Random Forest Classifier

Highest accuracy score: 0.86

#### Reasoning for testing:

- Gathers forecasts from each tree and predicts based on majority of votes
- Meta-estimator, enhancing models accuracy and prevents over-fitting.

2nd

### Gradient Boosting Classifier

Highest accuracy score: 0.83

#### Reasoning for testing:

- Powerful boosting algorithm
- Updates weights based on gradients
- Good for classification improvement

3rd

#### **SVM**

Highest accuracy score: 0.82

#### Reasoning for testing:

- Mostly utilized for classification difficulties
- Creates three dimension classification model

### Machine learning model optimization

1st

### Random Forest Classifier

Highest accuracy score: 0.86

Optimized over three times

Final optimization highest when not using standard scaler (0.86 vs 0.82)

Chosen model for app

```
# Create a random forest classifier
rf_model2 = RandomForestClassifier(n_estimators=400, random_state=78)
```

```
Accuracy Score: 0.8206521739130435
Classification Report
                         recall f1-score support
             precision
                  0.74
                           0.88
                                     0.80
                                                 77
                  0.90
                           0.78
                                                107
                                     0.83
                                                184
                                     0.82
  macro avg
                  0.82
                           0.83
                                     0.82
                                                184
                  0.83
                           0.82
weighted avg
                                     0.82
```

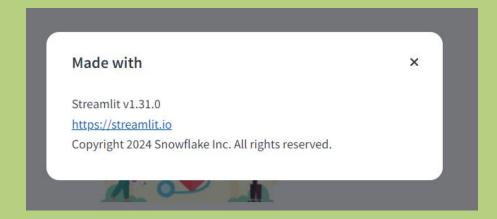
```
# Create a random forest classifier
rf_model = RandomForestClassifier(n_estimators=200, random_state=42, max_depth=10, min_samples_split=2, min_samples_leaf=1)
```

The same features were then ran again with not standard scaler for a third optimization and achieved a 0.86 accuracy score.

```
Accuracy Score: 0.8206521739130435
Classification Report
             precision
                         recall f1-score support
                  0.74
                           0.88
                                     0.80
                                                 77
                                     0.83
                                                107
                  0.90
                           0.78
                                     0.82
                                                184
   accuracy
  macro avg
                  0.82
                           0.83
                                     0.82
                                                184
weighted avg
                  0.83
                           0.82
                                     0.82
                                                184
```



#### **Streamlit**



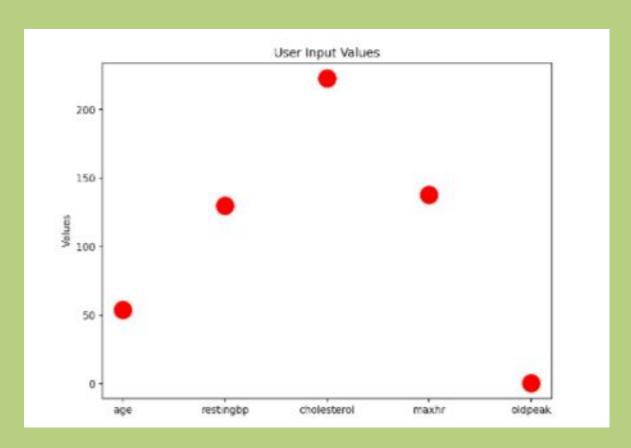
import streamlit as st install streamlit by running: pip install streamlit to run: streamlit run app.py

### Front-end features



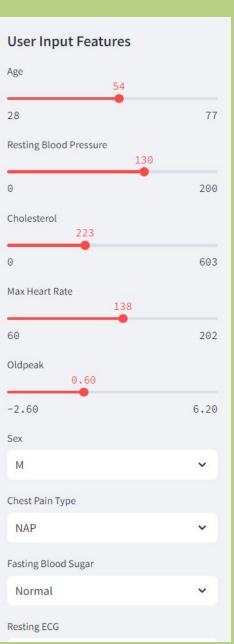


### Dynamic plotting and responsive design



# # Scatter plot for user input ax1.scatter(user\_input\_labels, user\_input\_values, color='red', marker='o', s=3\*100) # Adjust the size with s parameter ax1.set\_ylabel('Values') ax1.set\_title('User Input Values') # Bar chart for numeric features in the dataset numeric\_features = ['age', 'restingbp', 'cholesterol', 'maxhr', 'oldpeak'] ax2.barh(numeric\_features, heart\_data[numeric\_features].mean(), color='red') ax2.set\_xlabel('Mean Values') ax2.set\_title('Numeric Features in Dataset') st.pyplot(fig)

### Interactive sidebar for user interaction



# Al powered User Interaction





#### Lifestyle Advice:

To reduce the risk of heart disease, here are some general recommendations based on the given input:

- Maintain a healthy weight: It's important to keep your body weight within a healthy range by eating a balanced diet and engaging in regular physical activity.
- Monitor blood pressure: Keep track of your blood pressure levels and consult with a healthcare professional to manage and maintain healthy levels.
- Consider cholesterol levels: Pay attention to your cholesterol levels and make adjustments to your diet and lifestyle if needed to keep them within a healthy range.
- Manage stress: Incorporate stress-reducing activities such as meditation, deep breathing, or yoga to help manage stress levels.
- 5. Follow a healthy diet: Aim to consume a diet that is rich in fruits



Powered by OpenAl



Advices
Powered by
OpenAl API

#### **ChatGPT Response:**

Sure! Here are some of the best apps that can help you stay healthy and count your walking distance:

- Step Tracker: Step Tracker is a simple pedometer app that tracks your walking distance, steps taken, and calories burned. It's a great tool for tracking your daily physical activity.
- MapMyWalk: MapMyWalk uses GPS to track and map your walking route. It also provides audio feedback on your pace, distance, and calorie burn, making it a great



Lifestyle Advice from ChatGPT





Oser Questions
Answered by
ChatBOT

#### Chat with Hearty: Your AI Support Companion:

Hello! My name is Hearty, your dedicated support companion! Celebrate your journey towards transformative lifestyle changes and remarkable achievements. How may I assist you on this empowering path? I can provide personalized advice tailored just for you.

You:



Answered by ChatBOT



### **Live Demonstration**

#### **CardioGuard Al**

Predicting Heart Disease Risk & Providing Lifestyle Advice using Al

### Welcome to CardioGuard AI: Your Personalized Health Assistant!

Empower yourself with personalized predictions for heart disease risk and actionable lifestyle advice, all powered by cutting-edge AI technology.

To get started, simply enter your information in the sidebar, and let us guide you through insightful predictions and tailored recommendations.

Don't forget to engage with our friendly chatbot, Hearty, who's here to answer your questions and provide additional support along your health journey!



Healthy Hearts with CardioGuard Al



Life-Style Advices Powered by OpenALAPI



Health APPs by ChatGPT



User Questions Answered by ChatBOT



### Thanks!



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