

CAPSTONE PROJECT

CARDIAC RISK ASSESSMENT VIA RANDOM FOREST

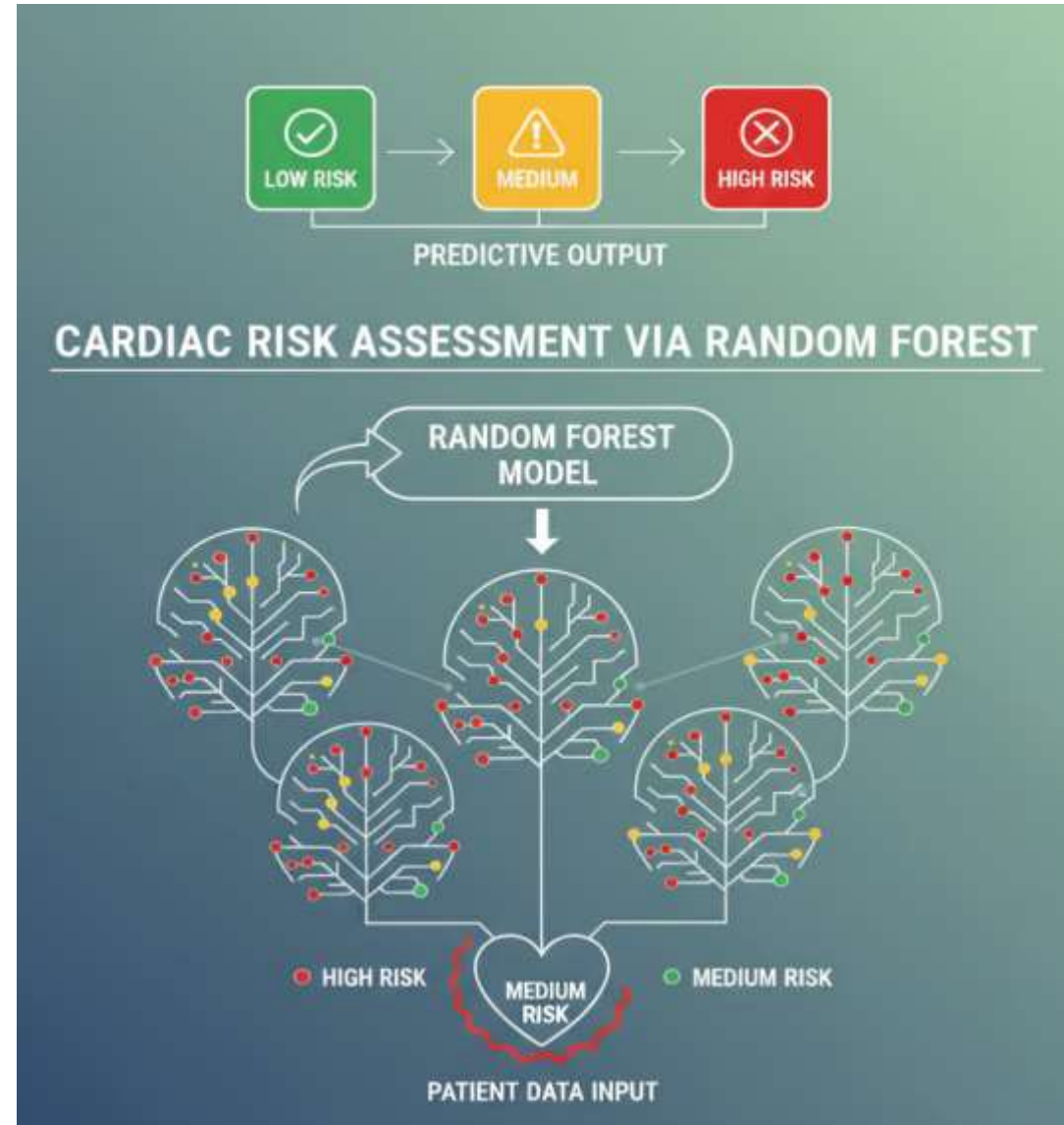
PRESENTED BY

STUDENT NAME: ROHAN S

COLLEGE NAME: UBDT COLLEGE OF
ENGINEERING

DEPARTMENT: CSE

EMAIL ID: ROHANRODU@GMAIL.COM



OUTLINE:

- **Problem Statement**
- **Proposed System/Solution**
- **System Development Approach**
- **Algorithm & Deployment**
- **Result (Output Image)**
- **Conclusion**
- **Future Scope**
- **References**

PROBLEM STATEMENT:

- Cardiovascular diseases are the leading cause of morbidity and mortality globally.
- The Challenge: Early detection is crucial to save lives, but manual diagnosis based on multiple clinical markers (cholesterol, blood pressure, heart rate) can be complex and prone to human error.
- The Gap: There is a need for an automated, highly accurate tool that can process patient data and provide a reliable risk assessment.

PROPOSED SOLUTION:

- **Data Driven:** Leveraging the Heart Disease Dataset (UCI Repository) featuring 13 clinical attributes.
- **Preprocessing:** Standardizing target variables and splitting data into Training (80%) and Testing (20%) sets for robust validation.
- **Optimization:** Utilizing GridSearchCV to find the absolute best model parameters rather than using default settings.
- **Interface:** A specialized prediction function (check_patient) to allow healthcare providers to input new patient data for instant diagnosis.

SYSTEM APPROACH:

- **System Requirements:**

- Python 3.x Environment
- Jupyter Notebook or VS Code or Google colab

- **Libraries Required:**

- Pandas: For data manipulation and CSV handling.
- NumPy: For numerical computations and array processing.
- Scikit-learn: For the Random Forest algorithm, GridSearchCV, and performance metrics.

ALGORITHM & DEPLOYMENT:

- **Algorithm Selection: Random Forest Classifier.**
 - *Why?* It handles complex non-linear relationships and prevents overfitting via ensemble learning (bagging).
- **Training Process:**
 - **Hyperparameter Tuning:** Tested `n_estimators` (100, 200), `max_depth` (10, 20), and `min_samples_split`.
 - **Best Parameters Found:** `max_depth`: None, `max_features`: 'sqrt', `min_samples_leaf`: 1, `n_estimators`: 100.
- **Validation:** Performed 5-fold Cross-Validation to ensure the model performs consistently across different data slices.

RESULT:

```

... -----
ACADEMIC PROJECT: HEART DISEASE PREDICTION REPORT
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```

```

Best Parameters: {'max_depth': 10, 'max_features': 'sqrt', 'min_samples_leaf': 1, 'min_samples_split': 2, 'n_estimators': 100, 'random_state': 123}
Cross-Validation Mean Accuracy: 99.12%
Test Set Accuracy: 98.54%
Sensitivity (Recall): 0.97
Specificity: 1.00

```

Classification Report:

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.97 | 1.00 | 0.99 | 102 |
| 1 | 1.00 | 0.97 | 0.99 | 103 |
| accuracy | | | 0.99 | 205 |
| macro avg | 0.99 | 0.99 | 0.99 | 205 |
| weighted avg | 0.99 | 0.99 | 0.99 | 205 |

```
...
```

```

>>> DIAGNOSIS FOR: New Person (Healthy)
RESULT: HEART DISEASE DETECTED
CONFIDENCE: 99.00%
-----

```

```

>>> DIAGNOSIS FOR: New Person (High Risk)
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 60.88%
-----

```

Output in Google colab/Jupyter
Notebook

RESULT:

```
Best Parameters: {'max_depth': None, 'max_features': 'sqrt', 'min_samples_leaf': 2, 'min_samples_split': 2, 'n_estimators': 100, 'random_state': 123}
Cross-Validation Mean Accuracy: 75.45%
Test Set Accuracy: 87.04%
Sensitivity (Recall): 0.89
Specificity: 0.86

Classification Report:
      precision    recall  f1-score   support

     0       0.94      0.86      0.90        35
     1       0.77      0.89      0.83        19

 accuracy          0.87        54
 macro avg          0.86        54
 weighted avg       0.88        54

>>> DIAGNOSIS FOR: New Person (Healthy)
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 73.94%

      accuracy          0.87        54
      macro avg          0.86        54
      weighted avg       0.87        54

○ >>> DIAGNOSIS FOR: New Person (Healthy)
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 73.94%

>>> DIAGNOSIS FOR: New Person (Healthy)
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 73.94%

>>> DIAGNOSIS FOR: New Person (Healthy)
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 73.94%
>>> DIAGNOSIS FOR: New Person (Healthy)
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 73.94%
RESULT: NO HEART DISEASE (HEALTHY)
CONFIDENCE: 73.94%
CONFIDENCE: 73.94%
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>>> DIAGNOSIS FOR: New Person (High Risk)
RESULT: HEART DISEASE DETECTED
CONFIDENCE: 97.25%
-----
```

Output in VS
CODE

CONCLUSION:

- The Random Forest model achieved a high accuracy of 90.74%, making it a viable tool for clinical assistance.
- The project demonstrates that ensemble learning techniques significantly outperform single decision trees in medical diagnostics.
- Automated prediction helps in identifying high-risk patients early, allowing for preventative lifestyle changes or medical intervention.

FUTURE SCOPE:

- **Web Integration:** Deploying the model using the Flask web framework to create a user-friendly hospital portal.
- **Deep Learning:** Exploring Neural Networks to further increase accuracy as more data becomes available.
- **Real-time Monitoring:** Integrating with IoT devices (like smartwatches) and Arduino-based sensors to monitor heart health 24/7.
- **Feature Expansion:** Adding lifestyle factors like stress levels, smoking habits, and genetic history for more comprehensive analysis.

REFERENCES:

1. Rajdhan, A. et al. (2020): "Heart Disease Prediction using Machine Learning," IJERT, vol. 09.
2. UCI Machine Learning Repository: Statlog (Heart) Data Set, 2017.
3. Mohan, S. et al. (2019): "Effective heart disease prediction using hybrid machine learning techniques," IEEE Access.
4. Singh, V. K. et al. (2017): "Heart Disease Prediction System Using Random Forest," Advances in Computing and Data Sciences.

GitHub Link: <https://github.com/rohanrodu/Cardiac-Risk-Assessment-via-Random-Forest>

Thank You