Group 3 Project Proposal

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Chosen Dataset

- Heart Failure Prediction
- The dataset contains 918 instances and 12 features.

Variables

- Age: Patient's age in years | Sex: Gender [M/F]
- ChestPainType: Type of chest pain [Typical Angina, Atypical Angina, Non-Anginal Pain, Asymptomatic]
- RestingBP: Resting blood pressure [mm Hg] | Cholesterol: Serum cholesterol [mm/dl]
- FastingBS: Fasting blood sugar [1: if FastingBS > 120 mg/dl, 0: otherwise]
- **RestingECG**: Resting electrocardiogram results [Normal, ST, LVH]
- MaxHR: Maximum heart rate achieved
- ExerciseAngina: Presence of exercise-induced angina [Y/N]
- Oldpeak: ST depression induced by exercise
- ST Slope: Slope of the peak exercise ST segment
- HeartDisease: Target variable indicating presence of heart disease [1: heart disease, 0: Normal]

Codes

<pre>import pandas as pd df = pd.read_csv("heart.csv") def resumetable(df): print(f"data shape: {df.shape}") summary = pd.DataFrame(df.dtypes, columns = ['data type']) summary = summary.reset_index() summary = summary.rename(columns = {"index": "feature"}) summary["Num_Null"] = df.isnull().sum().values summary["Num_Unique"] = df.loc[0].values summary["First_Value"] = df.loc[1].values summary["Third_Value"] = df.loc[2].values return summary resumetable(df)</pre>	data	a shape: (91 feature		Num_Null	Num_Unique	First_Value	Second_Value	Third_Value
	0	Age	int64	0	50	40	49	37
	1	Sex	object	0	2	М	F	М
	2	ChestPainType	object	0	4	ATA	NAP	ATA
	3	RestingBP	int64	0	67	140	160	130
	4	Cholesterol	int64	0	222	289	180	283
	5	FastingBS	int64	0	2	0	0	0
	6	RestingECG	object	0	3	Normal	Normal	ST
	7	MaxHR	int64	0	119	172	156	98
	8	ExerciseAngina	object	0	2	N	Ν	N
	9	Oldpeak	float64	0	53	0.0	1.0	0.0
	10	ST_Slope	object	0	3	Up	Flat	Up
	11	HeartDisease	int64	0	2	0	1	0

Question Ideas:

- Which clinical features are most predictive of heart disease?
- How does stress level impact the likelihood of developing a sleep disorder?
- Which classification algorithm yields the highest accuracy?
- How does feature engineering affect model performance for this dataset?

Methods

- Data Preprocessing
 - clean data by handling potential outliers
 - perform exploratory data analysis to understand variable distributions
 - Check for and address class imbalance of target feature
 - Encode categorical variables + Scale numerical features
- Feature Selection & Engineering
 - Apply correlation analysis to identify relationships between features
 - Use feature importance techniques to identify most predictive variables
 - Create interaction terms between important features
 - Conduct feature selection methods (SelectKBest)
- Model Development

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- Split data into training, validation, and test sets
 - Implement multiple classification algorithms
 - Logistic regression
 - Random Forest (Classifier)
 - Support Vector Machine (SVC)
 - Ensemble Model (Voting)
 - Conduct Hyperparameter Tuning using GridsearchCV + RandomSearchCV
- Threshold optimization for each model
- Model Evaluation
 Compa
 - Compare models using evaluation metrics
 - Accuracy, Precision, Recall, F1-Score, ROC Curve, and AUC