Salary Prediction using Ensemble Learning

This project demonstrates how to predict salaries based on features like experience and age using Ensemble Lear

Table of Contents:

- Overview
- Algorithms Used
- Dataset
- Project Structure
- Requirements
- How to Run
- Results
- License

Overview:

The goal of this project is to:

- Predict the salary of individuals based on features like experience and age.
- Use ensemble learning techniques to improve model performance.
- Compare individual models vs ensemble predictions.

Algorithms Used:

- Random Forest Regressor: A bagging-based ensemble method using multiple decision trees.
- Gradient Boosting Regressor: A boosting-based method that combines weak learners sequentially.
- Voting Regressor: Aggregates predictions from multiple models by averaging.

Dataset:

A sample synthetic dataset is used with the following features:

Experience (Years), Age, Salary (Target)

You can replace this with a real-world dataset (e.g., from Kaggle or job sites) for better prediction.

Project Structure:

salary-prediction-ensemble/

■■■ salary_prediction.py # Main Python script
■■■ README.md # Project readme file

■■■ requirements.txt # Python dependencies (optional)

Requirements:

Install dependencies using pip: pip install pandas scikit-learn

How to Run:

- 1. Clone this repository or copy the files.
- 2. Run the script:

python salary_prediction.py

- 3. The script will:
- Split the dataset
- Train individual models (Random Forest, Gradient Boosting)
- Combine them using Voting Regressor
- Print the Mean Squared Error (MSE) for each model

Results:

Example output:

Random Forest MSE: 1245.67 Gradient Boosting MSE: 1350.25 Voting Regressor MSE: 1180.42

This shows that the ensemble model performed slightly better in terms of lower error.

License:

This project is open-source and free to use for learning or personal use.