1. Frontend (React)

User Interface:

- Dataset Upload: Create a form where users can upload their datasets (e.g., CSV files).
- Algorithm Selection: Provide options for users to select from the available ML algorithms.
- Parameters Configuration: Allow users to configure hyperparameters for the selected algorithms.
- Results Display: Show the results of the ML model's predictions or evaluations.

Components:

- o **UploadForm:** Handles file uploads and sends the file to the backend.
- o **AlgorithmSelector:** Lets users choose from the available algorithms.
- o ParameterConfig: Allows users to set hyperparameters.
- ResultsDisplay: Shows the model's results or evaluation metrics.

API Calls:

Use fetch or Axios to send data to the Flask backend and receive results.

2. Backend (Flask + scikit-learn)

Endpoints:

- o **Upload Endpoint:** Receives and processes the uploaded dataset.
 - Parse the dataset and possibly store it temporarily or in a database.
- Train Model Endpoint: Accepts user-selected algorithm and parameters.
 - Load the dataset, train the selected model, and evaluate its performance.
- Predict Endpoint: Accepts new data and uses the trained model to make predictions.

Processing Flow:

Receive Dataset:

- Save the uploaded dataset.
- Perform initial checks (e.g., data format, completeness).

Algorithm Training:

- Based on user selection, instantiate the appropriate scikit-learn model.
- Configure the model with the provided hyperparameters.
- Train the model on the dataset.
- Evaluate the model and return performance metrics (e.g., accuracy, precision).

Prediction:

• Use the trained model to make predictions on new input data.

Integration:

- Use Flask routes to handle requests from the React frontend.
- o Ensure proper data validation and error handling.

3. ML Algorithms (Initial 5 Algorithms)

- Classification Algorithms:
 - Logistic Regression
 - Decision Tree Classifier
 - Random Forest Classifier
 - Support Vector Machine (SVM)
 - K-Nearest Neighbors (KNN)
- Considerations:
 - Each algorithm may have different hyperparameters that users can configure.
 - o Implement model training and evaluation for each selected algorithm.

4. Additional Features

- Model Persistence: Optionally save and load trained models for reuse.
- **Visualization:** Provide visualizations of results (e.g., confusion matrix, ROC curves).
- **Error Handling:** Implement robust error handling and user feedback mechanisms.

Deployment:

- Frontend: Build and deploy your React application.
- **Backend:** Deploy your Flask application (e.g., using Heroku, AWS, or another cloud service).

Workflow Example:

- 1. **User uploads a dataset** via the React frontend.
- 2. Frontend sends the dataset to the Flask backend.
- 3. **Flask saves the dataset** and returns a success message.
- 4. **User selects an algorithm** and configures parameters.
- 5. **Frontend sends the algorithm choice and parameters** to the Flask backend.
- 6. Flask trains the model using scikit-learn and returns evaluation metrics.
- 7. **Frontend displays the results** and possibly allows the user to make predictions with new data