

# Naive Bayes Model Integrated with Website using FastAPI

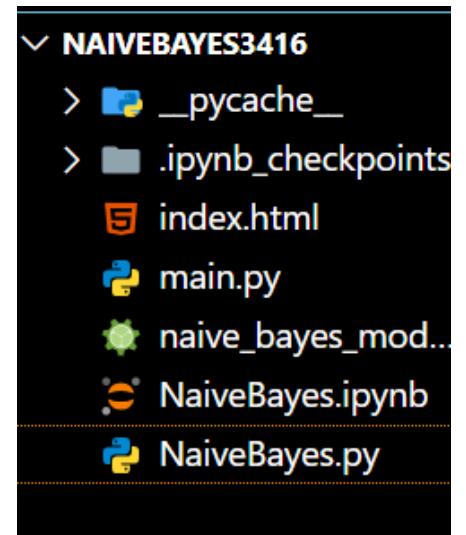
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## 1. Overview

This project implements a Naive Bayes classifier using the Adult Income dataset from OpenML. The model predicts whether an individual earns more than \$50K per year based on selected features.

### List of Files

1. **NaiveBayes.ipynb** - Jupyter Notebook for training and saving the model.
2. **main.py** - FastAPI backend to serve predictions.
3. **index.html** - Frontend with inline CSS & JavaScript for user interaction.
4. **naive\_bayes\_model.pkl** - Saved Naive Bayes model.
5. **NaiveBayes.py** - Python script version of the Jupyter Notebook.



## 2. Installation & Setup

### Prerequisites

Ensure you have Python installed, along with the following dependencies:

```
pip install fastapi uvicorn scikit-learn pandas numpy pydantic
```

### Running the FastAPI Server

Start the FastAPI backend with:

```
uvicorn main:app --host 127.0.0.1 --port 8011 --reload
```

### 3. Training the Model & Generating Pickle File

The Jupyter Notebook (`naive_bayes.ipynb`) loads the dataset, preprocesses it, trains a **Gaussian Naive Bayes** model, and saves it as a pickle file.

```
PS C:\Users\KIIT\Documents\AD22053416\SVM3416> jupyter nbconvert --to script  
NaiveBayes.ipynb
```

### 4. FastAPI Backend (`main.py`)

The backend loads the trained model and provides an API endpoint to receive feature inputs and return predictions.

```
@app.post("/predict")  
  
async def predict(data: InputData):  
  
    try:  
  
        features = np.array(data.features).reshape(1, -1)  
  
        prediction = model.predict(features)[0]  
  
        return {"prediction": int(prediction)}  
  
    except Exception as e:  
  
        raise HTTPException(status_code=400, detail=str(e))  
  
if __name__ == "__main__":  
  
    import uvicorn  
  
    uvicorn.run(app, host="127.0.0.1", port=8011)
```

## 5. Frontend (index.html) with Inline CSS & JavaScript

This file allows users to input data and get predictions from the FastAPI backend.

```
async function getPrediction() {
    let input = document.getElementById("features").value;
    let featureArray = input.split(",").map(Number);

    let response = await fetch("http://127.0.0.1:8011/predict", {
        method: "POST",
        headers: { "Content-Type": "application/json" },
        body: JSON.stringify({ features: featureArray })
    });

    let data = await response.json();
    let predictionText = data.prediction === 1 ? "Income > $50K" :
    "Income ≤ $50K";
    document.getElementById("result").innerText = "Prediction: " +
    predictionText;
}
```

## 6. Testing the Integration

1. **Run FastAPI Backend:**  
uvicorn main:app --host 127.0.0.1 --port 8011 --reload
2. **Open index.html in a browser.**
3. **Enter sample values** such as:  
39, 13, 2174, 0, 40
4. **Click Predict** → The output will show either:
  - Prediction: Income > \$50K
  - Prediction: Income ≤ \$50K

## Naive Bayes Income Predictor

This model predicts whether a person earns more than \$50K based on input features.

Enter the following values :

1. Age, 2. Education-num, 3. Capital-gain, 4. Capital-loss, 5. Hours-per-week

39, 13, 2174, 0, 40

Predict

Prediction: Income  $\leq$  \$50K

## Naive Bayes Income Predictor

This model predicts whether a person earns more than \$50K based on input features.

Enter the following values :

1. Age, 2. Education-num, 3. Capital-gain, 4. Capital-loss, 5. Hours-per-week

28, 16, 5000, 0, 45

Predict

Prediction: Income  $>$  \$50K

## 7. Conclusion

This project successfully integrates a Naive Bayes classifier with a FastAPI backend and an HTML frontend for user interaction. Users can input features and receive predictions on whether they earn more than \$50K. This setup can be extended to other classification problems with minor modifications.