

Sales Prediction Pipeline Documentation

Introduction

This documentation provides a comprehensive guide to the Sales Prediction Pipeline. The project predicts daily sales for retail stores using machine learning and deep learning models. It also includes a REST API for real-time predictions, making it suitable for deployment in business environments.

System Overview

The pipeline consists of the following components:

- Data Preprocessing:**
 - Handling missing data.
 - Normalization and feature engineering.
 - Preparing input features for training.
 - Model Training:**
 - Machine Learning Models:** Random Forest and XGBoost.
 - Deep Learning Models:** Feedforward Neural Network.
 - Model Deployment:**
 - REST API built with FastAPI.
 - Dockerized for easy deployment on cloud platforms.
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Prerequisites

Before starting, ensure the following:

- Python:** Version 3.8 or higher.
- Basic understanding of machine learning and REST APIs.
- Installed dependencies from `requirements.txt`.

Setup Instructions

1. Clone the Repository

```
git clone https://github.com/your-username/sales-prediction-pipeline.git  
cd sales-prediction-pipeline
```

2. Create a Virtual Environment

```
python -m venv venv  
source venv/bin/activate # On Windows: venv\Scripts\activate
```

3. Install Dependencies

```
pip install -r requirements.txt
```

4. Prepare the Dataset

Place the dataset (`data.zip`) in the `data/` directory.

Pipeline Execution

1. Data Preprocessing

Run the preprocessing script to clean and prepare the data:

```
python scripts/pipeline.py
```

2. Model Training

The training process saves the trained model in the `models/` directory.

3. API Launch

Run the REST API for predictions:

```
uvicorn main:app --reload
```

REST API Usage

Endpoints

- **GET /**: Health check endpoint to verify the API is running.
- **POST /predict/**: Accepts input data for prediction.

Example **POST** Request

```
{  
  "CompetitionDistance": 500,  
  "StoreType": "a",  
  "DayOfWeek": 3  
}
```

Example Response

```
{  
  "prediction": 4321.23  
}
```

Technical Details

Deep Learning Model

- **Architecture:**
 - Input layer with normalized features.
 - Hidden layers with ReLU activations and dropout.
 - Output layer for regression.
- **Loss Function:** Mean Squared Error (MSE).
- **Optimizer:** Adam.

Training Metrics

- **Training Loss:** 0.23
 - **Validation Loss:** 0.27
 - **Accuracy:** 91%
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Deployment

Local Deployment

Run the following command:

```
uvicorn main:app --host 0.0.0.0 --port 8000
```

Docker Deployment

Create a Dockerfile:

```
FROM python:3.9
```

```
WORKDIR /app
```

```
COPY . /app
```

```
RUN pip install -r requirements.txt
```

```
CMD ["uvicorn", "main:app", "--host", "0.0.0.0", "--port", "8000"]
```

1.

Build and Run Docker Image:

```
docker build -t sales-prediction-api .
```

```
docker run -p 8000:8000 sales-prediction-api
```

2.

Cloud Deployment

Use platforms like AWS, Heroku, or Azure. Refer to their respective documentation for deployment steps.

Repository Structure

```
sales-prediction-pipeline/
```

```
|
```

```
|— data/          # Data folder
```

```
|— scripts/       # Scripts for preprocessing, training, and API
```

```
| |— data_preprocessing.py
```

```
| |— model_training.py
```

```
| |— pipeline.py
```

```
|— models/        # Saved models
```

```
|  └— model.pkl
```

```
|— main.py        # FastAPI application
```

```
|— requirements.txt # Dependencies
```

```
└— README.md      # Project documentation
```

Contributing

We welcome contributions!

- **Submit issues** for bugs or feature requests.
- **Create pull requests** to propose changes.

Contact Information

- **Author:** Amanuel Legesse
 - **Email:** ese.amanuel.legesse@gmail.com
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