Low Level Design (LLD)

FLIGHT FARE PREDICTION

Revision Number – 1.3 Last Date of Revision – 15-Jan-2023 Lally

Document Version Control:

Date	Version	Description	Author
14-01-2023	1.0	Abstract, Introduction Architecture	Lally
15-01-2023	1.1	Data Preprocessing	Lally

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Abstract

The recent international things had a large impact on the aviation sector because of several reasons. This impact has 2 class folks, the primary is business perspective and therefore the second is that the customers perspective. As safety is that the major reason for such impact on the aviation sector, the governments round the world amended totally different rules to their various airlines firms. These restrictions had created the supply of the flights and their attendant capability less. Taking of these factors in thought the value of the flight tickets has accrued and vary from one place to the opposite. Booking a flight price tag has split into 2, one is that the on-line and therefore the alternative is that the offline bookings. Each these have their various criteria for value of the price tag, one such example is that the server load and therefore the range of booking requests. during this machine learning implementation, we are going to see numerous factors that impact worth of the flight ticket price and predict the acceptable price of the ticket.

1. Introduction

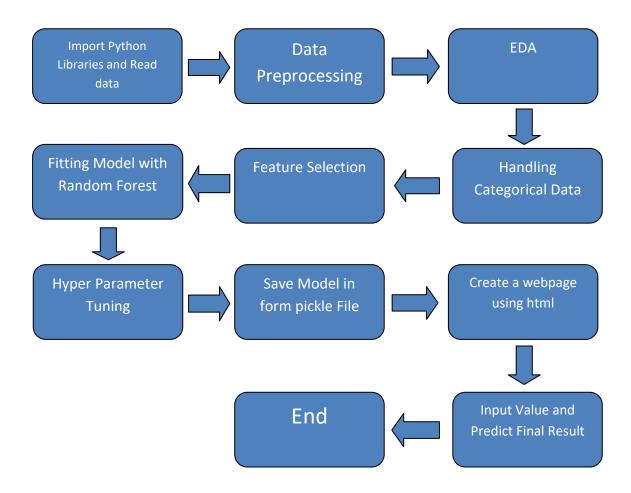
1.1. What is Low-Level design document?

The main purpose of this LLD documentation is to feature the required details of the project and supply the outline of the machine learning model and also the written code. This additionally provides the careful description on however the complete project has been designed end-to-end.

1.2. Scope

Low-level design (LLD) is a component-level design process that follows a step by step refinement process. This process can be used for designing data structures, required software architecture, source code and ultimately, performance algorithms. Overall, the data organization may be defined during requirement analysis and then refined during data design work

2. Architecture



3. Architecture Description

This project is to make associate interface for the user to grasp their approximate flight price ticket worth, additionally to the present, in would like of obtaining the important time project expertise we have a tendency to square measure mercantilism the gathered information into our own information then begin the project from the scratch.

3.1. Data Gathering

The data for the current project is being gathered from Kaggle dataset, the link to the data is

https://www.kaggle.com/datasets/nikhilmittal/flight-fare-prediction-mh

3.2. Data Description

There are about 10k+ records of flight information such as airlines, data of journey, source, destination, departure time, arrival time, duration, total stops, additional information, and price. A glance of the dataset is shown below.

Airline	e_of_Jour	Source	Destination	Route	Dep_	Time rrival_	Tim Duration	Total_Stop	ditional_Ir	Price
IndiGo	24/03/201	Banglore	New Delhi	BLR → DE	EL 22:20	01:10	22 N 2h 50m	non-stop	No info	3897
Air India	1/05/2019	Kolkata	Banglore	CCU → I>	(F05:50	13:15	7h 25m	2 stops	No info	7662
Jet Airway	9/06/2019	Delhi	Cochin	DEL -> LK	C 09:25	5 04:25	10 Ji 19h	2 stops	No info	13882
IndiGo	12/05/201	Kolkata	Banglore	CCU → N	A 18:05	5 23:30	5h 25m	1 stop	No info	6218
IndiGo	01/03/201	Banglore	New Delhi	BLR → N	A 16:50	21:35	4h 45m	1 stop	No info	13302
SpiceJet	24/06/201	Kolkata	Banglore	CCU → B	109:00	11:25	2h 25m	non-stop	No info	3873
Jet Airway	12/03/201	Banglore	New Delhi	BLR → BC	0118:55	5 10:25	13 N 15h 30m	1 stop	In-flight m	11087
Jet Airway	01/03/201	Banglore	New Delhi	BLR → BC	00:8010	05:05	02 N 21h 5m	1 stop	No info	22270
Jet Airway	12/03/201	Banglore	New Delhi	BLR → BC	0108:55	5 10:25	13 N 25h 30m	1 stop	In-flight m	11087
Multiple c	27/05/201	Delhi	Cochin	DEL → BO	0 11:25	5 19:15	7h 50m	1 stop	No info	8625
Air India	1/06/2019	Delhi	Cochin	DEL -> BI	F09:45	23:00	13h 15m	1 stop	No info	8907
IndiGo	18/04/201	Kolkata	Banglore	CCU → B	LI 20:20	22:55	2h 35m	non-stop	No info	4174
Air India	24/06/201	Chennai	Kolkata	MAA -> ((11:40	13:55	2h 15m	non-stop	No info	4667
Jet Airway	9/05/2019	Kolkata	Banglore	CCU → B	021:10	09:20	10 N 12h 10m	1 stop	In-flight m	9663
IndiGo	24/04/201	Kolkata	Banglore	CCU → B	LI 17:15	5 19:50	2h 35m	non-stop	No info	4804
Air India	3/03/2019	Delhi	Cochin	DEL -> AI	M16:40	19:15	04 N 26h 35m	2 stops	No info	14011
SpiceJet	15/04/201	Delhi	Cochin	DEL -> PI	N:08:45	5 13:15	4h 30m	1 stop	No info	5830
Jet Airway	12/06/201	Delhi	Cochin	DEL -> BO	0114:00	12:35	13 Ji 22h 35m	1 stop	In-flight m	10262

3.3. Tool Used

- Python 3.9 is employed because the programming language and frame works like numpy, pandas, sklearn and alternative modules for building the model.
- PyCharm is employed as IDE.
- For visualizations seaborn and components of matplotlib are getting used
- For information assortment prophetess info is getting used version management.

3.4. Data Pre-processing

Steps performed in pre-processing are:

- First the info sorts square measure being checked and located solely the value column is of sort number.
- Checked for null values as there square measure few null values, those rows square measure born.
- Converted all the desired column into the date time format.
- Performed one-hot cryptography for the desired columns.
- Scaling is performed for needed information.
- And, the info is prepared for passing to the machine learning formula

3.5. Model Building

The pre-processed information is then envisioned and every one the specified insights are being drawn, though from the drawn insights, the info is at random unfold however still modeling is performed with completely different machine learning algorithms to form positive we tend to cowl all the chances. And eventually, for sure random forest regression performed well and any hyper parameter calibration is finished to extend the model's accuracy.

3.6. Data from User

The data from the user is retrieved from the created HTML web page.

3.7. Data Validation

The data provided by the user is then being processed by app.py file and validated. The validated data is then sent for the prediction.

3.8. Rendering Result

The data sent for the prediction is then rendered to the web page.