Assignment 2

By: Abdulwahab Hawsawi & Abdulwahed Dahan

The input data for the GAI in details (Attach the data file).

The input data for the GAI is a the same data that would be used for the machine learning algorithm. It is historical data of some of the trips that were taken using the Wusool service from 2018 to 2023. The total count of the records is around 600,000.

The data was acquired from a website called OpenData: a Saudi website where different government agencies and organizations can publicly release some the data they collect. The dataset that will be used in the project comes The Human Resources Development Fund. The data is from 2018 to 2023. Each of the six years is associated with a CSV file that contains around 100,000 rows. While some of the rows gave odd numbers, e.g. the price of the trip is 0, these were very few.

The record files have a number of columns, 12 to be exact, but the program will use four of them: pickup longitude and latitude, and drop-off longitude and latitude.

Pickup

Longtitude

Pickup

Drop-off

Explain in details your GAI website with screenshots.

Our app is a python script that opens the files for all the records, take the relevant columns from every file, combines the data from all the records, normalize them, splits the data into training and testing dataset, then feeds it to the Generative Artificial Intelligence and machine learning algorithm. Then, the testing data will be used on both to see how accurate both models are within a specified range.

The model used for the machine learning algorithm is Multi-Layer Perceptron Regressor, which is a collection of perceptrons organized in layers between the input and output. The default settings will be used. The ones that are of interest are:

hidden_layer_size = 100

 $max_iter = 200$

learning_rate_init = 0.001

hidden layer sizes=100

40.40	04.00	04.00	40.40	45.04	40.00	
40.43	21.28	21.32	40.43	15.21	18.26	20.90
50.03	26.41	26.41	50.11	19.84	21.82	44.20
39.18	21.54	21.52	39.16	17.62	19.14	24.67
50.08	26.36	26.33	50.18	26.70	34.71	64.21
39.52	24.45	24.45	39.59	21.53	25.99	46.14
/lib/python3.	11/site-packag	es/sklearn/neura	l_network/_multi	.layer_perceptron.	py:1623: DataConver	sionWarning: A c
y = column_	or_1d(y, warn=	True)	l_network/_multi	.layer_perceptron.	py:1623: DataConver	sionWarning: A co
y = column_ Iteration 1,	or_1d(y, warn= loss = 145.522	True) 46511	il_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A co
y = column_ Iteration 1, Iteration 2,	or_1d(y, warn= loss = 145.522 loss = 134.489	True) 46511 02286	il_network/_multi	layer_perceptron.	py:1623: DataConver	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3,	or_1d(y, warn= loss = 145.522 loss = 134.489 loss = 133.928	True) 46511 02286 54396	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A c
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4,	or_1d(y, warn= loss = 145.522 loss = 134.489 loss = 133.928 loss = 133.540	True) 46511 02286 54396 11316	il_network/_multi	layer_perceptron.	py:1623: DataConver	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 5,	or_1d(y, warn= loss = 145.522 loss = 134.489 loss = 133.928 loss = 133.540 loss = 131.541	True) 46511 02286 54396 11316 43945	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 5, Iteration 6,	or_1d(y, warn= loss = 145.522 loss = 134.489 loss = 133.928 loss = 133.540 loss = 131.541 loss = 128.200	True) 46511 02286 54396 11316 43945 57647	l_network∕_multi	layer_perceptron.	oy:1623: DataConver:	sionWarning: A c
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 5, Iteration 6, Iteration 7,	or_1d(y, warn= loss = 145.522 loss = 134.489 loss = 133.928 loss = 133.540 loss = 131.541 loss = 128.200 loss = 124.094	True) 46511 02286 54396 11316 43945 57647 02550	il_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 5, Iteration 6, Iteration 7, Iteration 8,	or_1d(y, warn= loss = 145.522 loss = 134.489 loss = 133.928 loss = 133.540 loss = 131.541 loss = 128.200 loss = 124.094 loss = 116.369	True) 46511 02286 54396 11316 43945 57647 02550 33282	l_network/_multi	layer_perceptron.	py:1623: DataConver	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 5, Iteration 6, Iteration 7, Iteration 8, Iteration 9,	or_1d(y, warn=' loss = 145.522: loss = 134.489 loss = 133.540: loss = 131.541: loss = 128.200! loss = 126.4094 loss = 116.369: loss = 106.487	True) 46511 602286 524396 11316 43945 57647 602550 33282	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 6, Iteration 7, Iteration 8, Iteration 9, Iteration 10,	or_1d(y, warn=' loss = 145.522: loss = 134.489 loss = 133.540: loss = 131.541: loss = 128.200: loss = 124.694! loss = 116.369: loss = 196.487! loss = 96.604	True) 46511 602286 613316 43945 57647 602550 333282 90808	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 5, Iteration 6, Iteration 7, Iteration 8, Iteration 10, Iteration 10, Iteration 11,	or_1d(y, warn= loss = 145.522: loss = 133.489 loss = 133.540: loss = 131.541: loss = 128.200: loss = 124.094! loss = 16.369: loss = 96.604! loss = 88.548!	True) 46511 902286 54396 11316 43945 57647 902550 33282 90808	il_network∕_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A co
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 5, Iteration 7, Iteration 7, Iteration 9, Iteration 10, Iteration 11, Iteration 11,	or_1d(y, warn= loss = 145.522- loss = 134.489 loss = 133.540- loss = 131.541- loss = 128.200- loss = 124.994- loss = 166.487- loss = 96.604- loss = 82.404-	True) 46511 46511 54396 11316 43945 57647 82550 33282 90808 94847 99777	l_network/_multi	layer_perceptron.	py:1623: DataConver	sionWarning: A c
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 6, Iteration 6, Iteration 7, Iteration 8, Iteration 10, Iteration 11, Iteration 12, Iteration 12, Iteration 12, Iteration 13,	or_1d(y, warn= loss = 145.522. loss = 134.489 loss = 133.928 loss = 133.540 loss = 131.541 loss = 128.200 loss = 124.094 loss = 166.487 loss = 96.604 loss = 88.548 loss = 82.4044 loss = 76.950	True) 46511 62526 54396 11316 43945 57647 62550 33282 99808 94847 89777	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A c
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 6, Iteration 6, Iteration 8, Iteration 10, Iteration 11, Iteration 12, Iteration 12, Iteration 14,	or_1d(y, warn= loss = 145.522. loss = 134.4899 loss = 133.44899 loss = 133.540 loss = 131.541 loss = 128.200 loss = 124.094 loss = 16.369 loss = 16.487 loss = 96.604 loss = 82.4404 loss = 76.950 loss = 76.950 loss = 77.237	True) 46511 46511 54286 54396 11316 43945 57647 90259 33282 90888 94447 909777 96499 89384	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A c
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 4, Iteration 6, Iteration 7, Iteration 8, Iteration 10, Iteration 11, Iteration 12, Iteration 13, Iteration 13, Iteration 14, Iteration 15,	or_ld(y, warn= loss = 145.522 loss = 134.489 loss = 133.928 loss = 133.541 loss = 131.541 loss = 128.200 loss = 126.000 loss = 124.000 loss = 166.487 loss = 88.5484 loss = 82.404 loss = 76.950 loss = 76.950 loss = 72.397	True) 46511 402286 54396 11316 43945 57647 20259 33282 99888 994847 499777 966499 89384 30395	il_network∕_multi	layer_perceptron.	oy:1623: DataConver:	sionWarning: A c
y = column_ Iteration 1, Iteration 2, Iteration 3, Iteration 3, Iteration 5, Iteration 6, Iteration 7, Iteration 8, Iteration 10, Iteration 10, Iteration 11, Iteration 12, Iteration 13, Iteration 14, Iteration 16, Iteration 16, Iteration 16, Iteration 16, Iteration 17, Iteration 17, Iteration 18, Iteration 16, Iteration 16, Iteration 16,	or_1d(y, warn= loss = 145.522. loss = 134.4899 loss = 133.44899 loss = 133.540 loss = 131.541 loss = 128.200 loss = 124.094 loss = 16.369 loss = 16.487 loss = 96.604 loss = 82.4404 loss = 76.950 loss = 76.950 loss = 77.237	True) 66511 802286 54396 11316 43945 57647 802559 93828 94847 99777 964499 893394 893395	l_network/_multi	layer_perceptron.	py:1623: DataConver:	sionWarning: A c

Machine

Prediction

ChatGPT

Prediction

Actual

Explain in details the output of your GAI website with screenshot of your results.

The GAI was supposed to take the data, pretend it is a machine learning algorithm, train on it, then take the testing data and test itself and give an accuracy score.

We managed to upload part of the data to LangChain in the form of a CSV file.

While there were ways to upload the CSV files, the model would not know how to train and test on it.

Explain how Machine Learning being used in your project with input data.

This step is done by first passing the training features, xtrain, and the training target, ytrain. Then, the test features, xtest std, are passed to ... test if the model has been trained correctly.

To test the accuracy, the predicted value is divided by the actual value. This will give us, percentage wise, how close the values are. It will then be decided whether or not the prediction is accurate based on the tolerance level. For example, if the tolerance level is set to 0.9, then an "accurate" result will be within 10% of the true value. 0.8 means within 20% or the true value.

Using the default values, and a tolerance value of 0.8, the training finished in 2 minutes and 48 seconds. The accuracy was 60.04%

Experimentation

Changing the default configurations yielded the following results:

- Increasing learning_rate_init from 0.001 to 0.01 increased the accuracy to 41.92%
- Reducing hidden_layer_sizes from 100 to 9, while keeping learning_rate_init at 0.01, increased the accuracy to 51.99%. In addition to the accuracy increase, the speed of the

training has been reduced to 32 seconds. However, this only happened twice. Afterwards, every subsequent try stuck around the high-30s in terms of accuracy.

- Returning to the default made the accuracy around 48%.
- Removing the 2023 dataset increased the accuracy to around 62%! It also improved the training time by reducing it to 1 minute and 11 seconds.
- Removing the 2023, 2022, 2021, and 2020 dataset reduced the accuracy to around 52%.

It seems the 2023 dataset a lot of datapoints that cause the model to learn bad information. The best accuracy is ensured when it is removed.

Here is an example of the correct predictions the model made, within 20% of the true value:

```
predicted: 16.588202975272583
                                  Actual: [19.58]
predicted: 19.283030191163917
                                  Actual: [21.03]
predicted: 12.477689107341178
                                  Actual: [13.86]
predicted: 33.008337765388745
                                  Actual: [36.85]
                                  Actual: [14.36]
predicted: 13.713172506810213
predicted: 17.816510138561256
                                  Actual: [17.99]
predicted: 41.1749594935717
                                Actual: [47.5]
predicted: 12.569987010028548
                                  Actual: [12.83]
predicted: 18.803941741631622
                                  Actual: [21.53]
predicted: 27.954936273942362
                                  Actual: [31.47]
predicted: 19.39791361048876
                                 Actual: [23.87]
predicted: 56.79312560781528
                                 Actual: [57.28]
predicted: 12.043761513819394
                                  Actual: [14.84]
predicted: 25.318241351192427
                                  Actual: [30.97]
predicted: 36.06199556468952
                                 Actual: [36.2]
predicted: 44.66638947478601
                                 Actual: [53.71]
predicted: 18.3545418828127
                                Actual: [19.98]
predicted: 44.03601936627344
                                 Actual: [53.51]
predicted: 17.928681381734116
                                  Actual: [18.5]
predicted: 11.09421141803097
                                 Actual: [12.5]
predicted: 31.864427695440742
                                  Actual: [34.86]
predicted: 45.23347002466219
                                 Actual: [53.5]
predicted: 18.48219847887576
                                 Actual: [22.08]
predicted: 55.34303045032199
                                 Actual: [61.77]
predicted: 35.538979929053234
                                  Actual: [41.91]
```

Here is an example of wrong predations the model made, outside the 20% threshold:

```
predicted: 14.891993882367721
                                  Actual: [10.29]
predicted: 16.543319829980707
                                  Actual: [11.74]
                                 Actual: [32.42]
predicted: 25.66390355747419
predicted: 12.021776819200028
                                  Actual: [29.02]
predicted: 16.080404686506274
                                  Actual: [25.69]
                                  Actual: [10.4]
predicted: 16.845504158069552
predicted: 46.814253557478985
                                  Actual: [65.35]
predicted: 45.65399081722004
                                 Actual: [28.72]
predicted: 21.94609959890758
                                 Actual: [28.04]
predicted: 19.18207195507338
                                 Actual: [12.62]
predicted: 19.89415920682546
                                 Actual: [15.98]
predicted: 29.794461985282158
                                  Actual: [20.83]
predicted: 27.63512063252011
                                 Actual: [20.35]
predicted: 31.10180645881539
                                 Actual: [23.55]
predicted: 21.30391623409994
                                 Actual: [14.92]
                                 Actual: [10.06]
predicted: 17.18945103520575
predicted: 22.42646289916289
                                 Actual: [17.58]
predicted: 52.46125661285843
                                 Actual: [42.37]
```

When reducing the threshold to within 10%, the accuracy got reduced to 37.41%. Here is some correct predictions:

```
predicted: 31.395891433741085
                                 Actual: [28.87]
predicted: 43.78802753134829 Actual: [40.43]
predicted: 19.283030191163917 Actual: [21.03]
predicted: 14.700580474193039 Actual: [13.41]
predicted: 13.198462354363143 Actual: [12.02]
predicted: 12.477689107341178 Actual: [13.86]
                                Actual: [32.43]
predicted: 32.990348785457805
predicted: 21.27987671952219 Actual: [20.26] predicted: 28.759570480971977 Actual: [27.52]
predicted: 13.713172506810213 Actual: [14.36]
predicted: 17.816510138561256 Actual: [17.99]
predicted: 12.569987010028548 Actual: [12.83]
predicted: 31.928337888766627 Actual: [30.9]
predicted: 30.419756331913785 Actual: [29.52]
predicted: 18.369481394272047
                                Actual: [16.99]
predicted: 56.79312560781528 Actual: [57.28] Actual: [26.71]
predicted: 43.209470174047986 Actual: [41.48]
predicted: 23.304870047361096 Actual: [23.22]
predicted: 27.357784584000992 Actual: [25.9]
```

The Machine Learning tools used in the project.

• Sckkit-learn: this library provided to model to be used in the app.

· Pandas: to read the CSV files

OpenAl and OS: to use chatgpt