

CPSC 531- Advance Database Management System

Final Project Airline Analysis

Web URL: http://airlineanalysis.epizy.com/

GitHub URL:

https://github.com/LencyLakhani/advancedatabase

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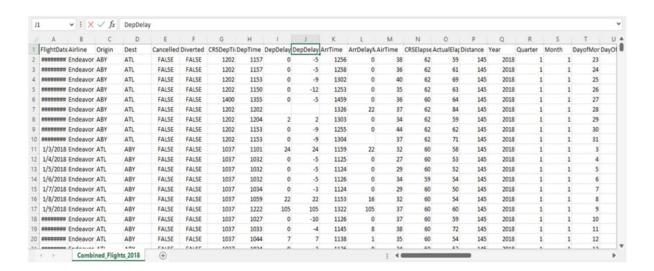
Problem Statement

Nowadays, there are many reasons for the delay and cancellation of flights. The <u>Federal Aviation Administration</u> estimates flight delays cost airlines \$22 billion yearly in the United States. Flight delays are inconvenient for passengers as well. The airline analysis allows for comparisons between different airlines, their time, arrival, and departure.

We analysed an airline dataset from 2018 to identify flights that were delayed for over 60 minutes, had a minimum airtime of 100 minutes, and examined their destinations. Our findings provide detailed information on specific airline flights that meet these criteria.

Dataset

We utilised a 1.9 GB dataset from the Kaggel website consisting of 1048575 records and 61 title columns. After cleaning the source CSV file, we extracted only the necessary fields such as Airline, Origin, Destination, Airtime, and departure delay time for the final output CSV file following analysis.



Technology & Tools Used

Backend:

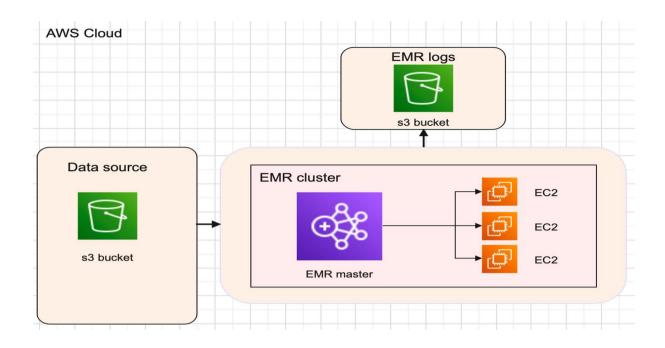
- AWS EMR (Amazon Elastic MapReduce)
- AWS S3 (Amazon Simple Storage Service)
- AWS EC2 (Amazon Elastic Compute Cloud)
- Apache Spark
- Python Pyspark SQL libraries

Frontend:

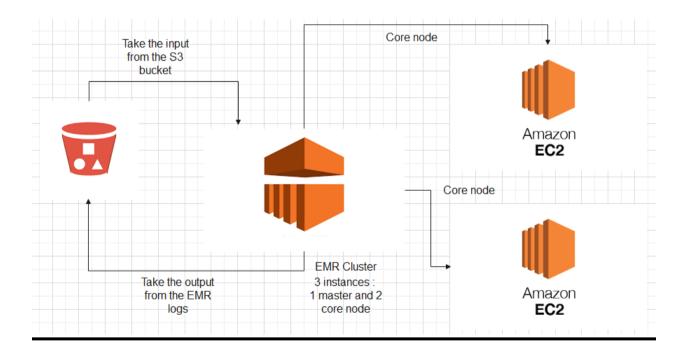
- HTML
- Javascript
- Bootstrap
- CSS

Overview Of Design

We used AWS EMR to handle large data servers on the cloud efficiently. This is possible because EMR comes equipped with pre-installed and pre-configured Spark and Hadoop. EMR necessitates worker nodes to execute the task, which we provided through AWS EC2 instances. To tackle a vast dataset, we carefully selected an EC2 instance with a configuration that offers 1 Master and 2 core nodes of m5.xlarge. Furthermore, the AWS S3 bucket served as our file system for storing project source, output, and Python pyspark files. Our project is represented in the diagram below, where EMR employs 3 instances of EC2 to execute the spark job. We accessed the source file through the S3 bucket and processed the data using the Python Pyspark file. EMR also generates logs in an S3 bucket.

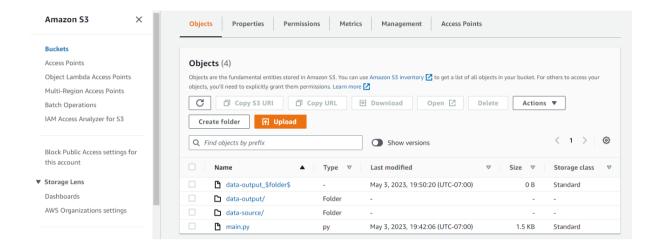


Architecture & Design

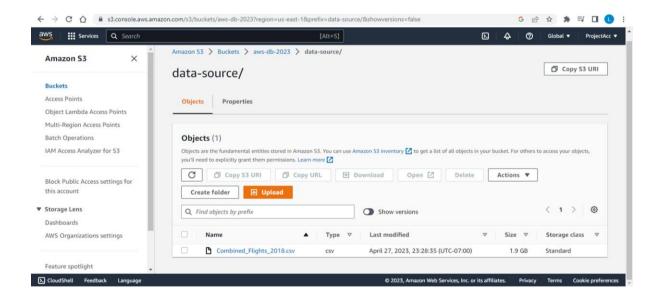


Implementation Overview

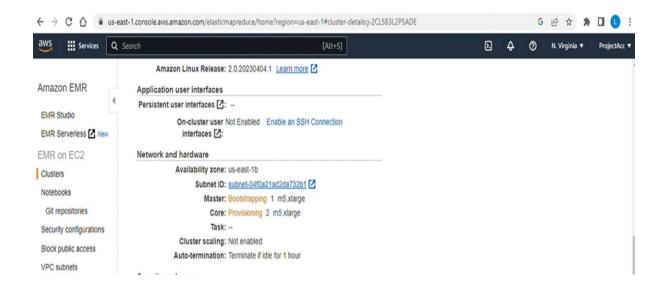
AWS S3 bucket console with data-source folder for storing airline source csv file, data-output folder to store the output csv file generated after the job completion of EMR cluster and the python file to perform spark job.



Uploaded the airline source CSV file, 1.9GB in size, to the data-source folder in S3 bucket.

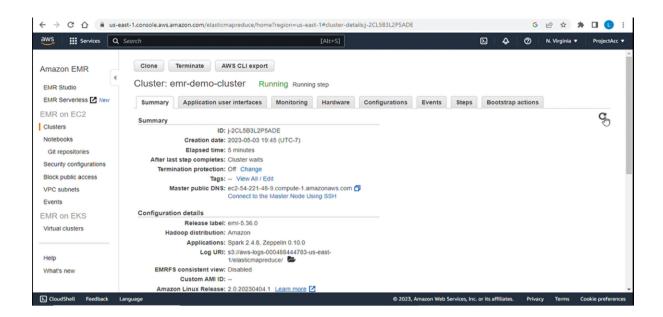


EMR cluster using three EC2 instances, consisting of one master node and two core nodes of m5.xlarge.

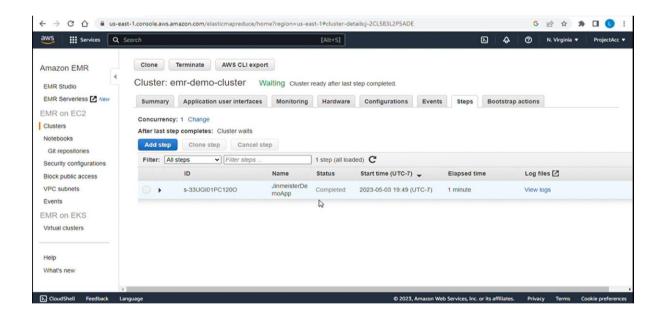


Python pyspark code to perform the spark job and process output data.

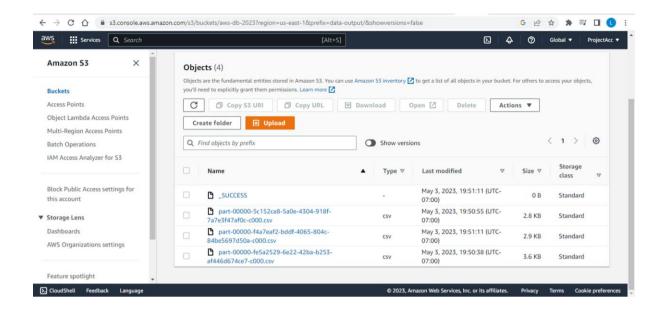
EMR running the python pyspark job file cluster instances to process the output.



EMR cluster ready with completed jb execution.

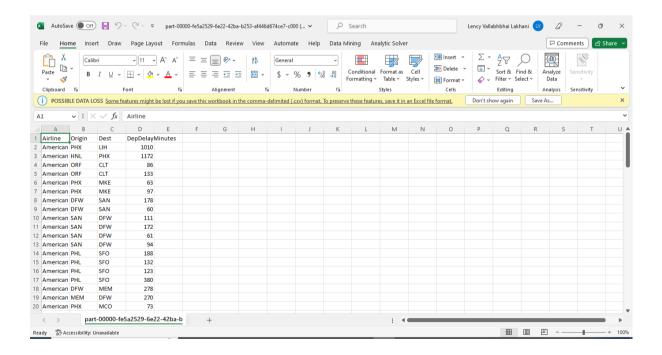


Output CSV files generated in the data-output folder of S3 bucket.



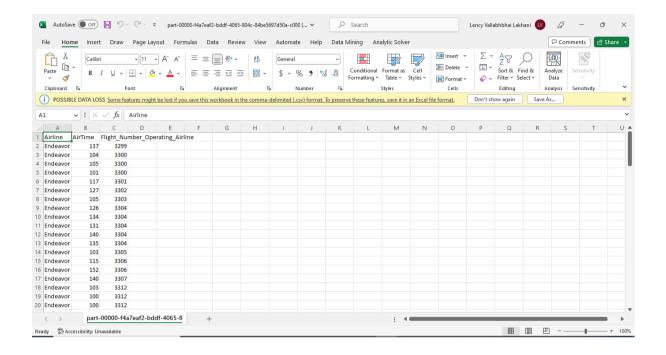
Output File 1: Delay Analysis Output for below query:

SELECT Airline,Origin,Dest,DepDelayMinutes FROM flights_dataset WHERE DepDelayMinutes >= 60 AND Airline = 'American Airlines Inc.' LIMIT 100



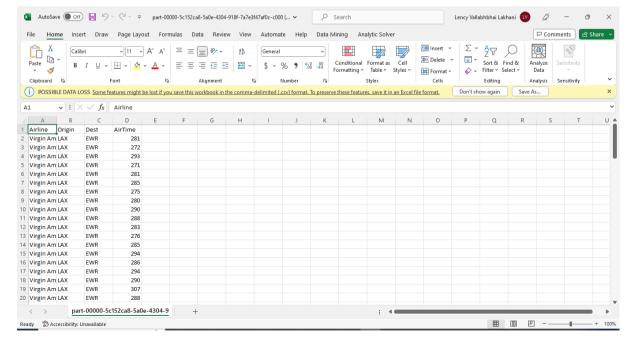
Output File 2: Airtime Analysis Output for below query:

SELECT Airline, AirTime, Flight_Number_Operating_Airline FROM flights_dataset WHERE AirTime >= 100 LIMIT 100

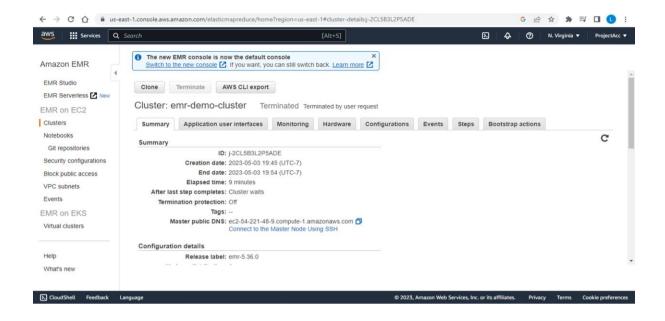


Output File 3: Destination Analysis Output for below query:

SELECT Airline,Origin,Dest,AirTime FROM flights_dataset WHERE Origin = 'LAX' AND Dest = 'EWR' ORDER BY Airline DESC LIMIT 100

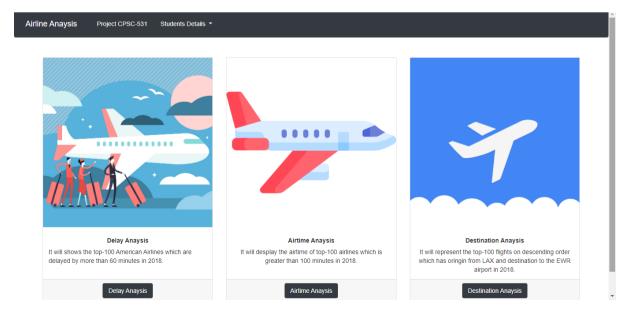


Terminate the cluster after job completion.



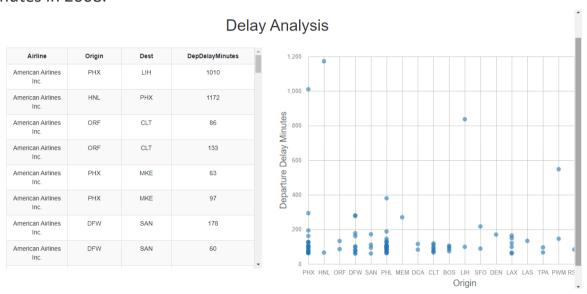
User Interface

We have created a user interface using javascript, HTML, CSS, and Bootstrap. This User interface consists of airline analysis which we have done using AWS services for 1.9 GB of airline dataset of 2008. It includes the three types of analysis "Delay analysis", "Airtime analysis" and "Destination Analysis". When the user clicks on each of the analysis, it will redirect to the page which will show the airline analysis and their related scatter graph.



1.Delay analysis

It will show the top-100 American Airlines which were delayed by more than 60 minutes in 2008.



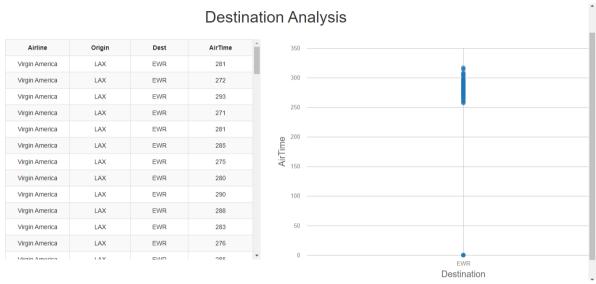
2.Airtime Analysis

It will display the airtime of top-100 airlines which is greater than 100 minutes in 2008.

		Airti	me	An	al	ysis	3															
Airline	AirTime	Flight_Number_Operating_Airline	A		200																_	
Endeavor Air Inc.	137.0	3299																				
Endeavor Air Inc.	104.0	3300																				
Endeavor Air Inc.	105.0	3300			180		T														\exists	
Endeavor Air Inc.	101.0	3300															1	3				
Endeavor Air Inc.	117.0	3301			160					-	-										-	
Endeavor Air Inc.	127.0	3302		AirTime								•										
Endeavor Air Inc.	105.0	3303																	•	,		
Endeavor Air Inc.	126.0	3304			140							1										
Endeavor Air Inc.	134.0	3304																				
Endeavor Air Inc.	131.0	3304			120		_			_											_	_
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Endower Air Inc	409 N	990E	*		100 32	99 330	330	1 3302	3303	3304 3	305 3	306 33	07 33	12 33	13 33	14 33	18 33	19 33	21 33	24 332	26 332	18 33
												Ai	rline	Flig	ht N	lumb	er					

3.Destination Analysis

It will represent the top-100 flights in descending order which has origin from LAX and destination to the EWR airport in 2018.



Step to run the code

- 1. Add the main.py file into the AWS s3 bucket.
- 2. In the EMR cluster, go to the steps.
- 3. In steps, follow the "add steps" button and then add the path of the main.py file which is stored in AWS s3 bucket.
- 4. Start the EMR cluster.
- 5. Waiting to the running step of the cluster.
- 6. Go to the s3 bucket output file.
- 7. In the last step, terminate the cluster.