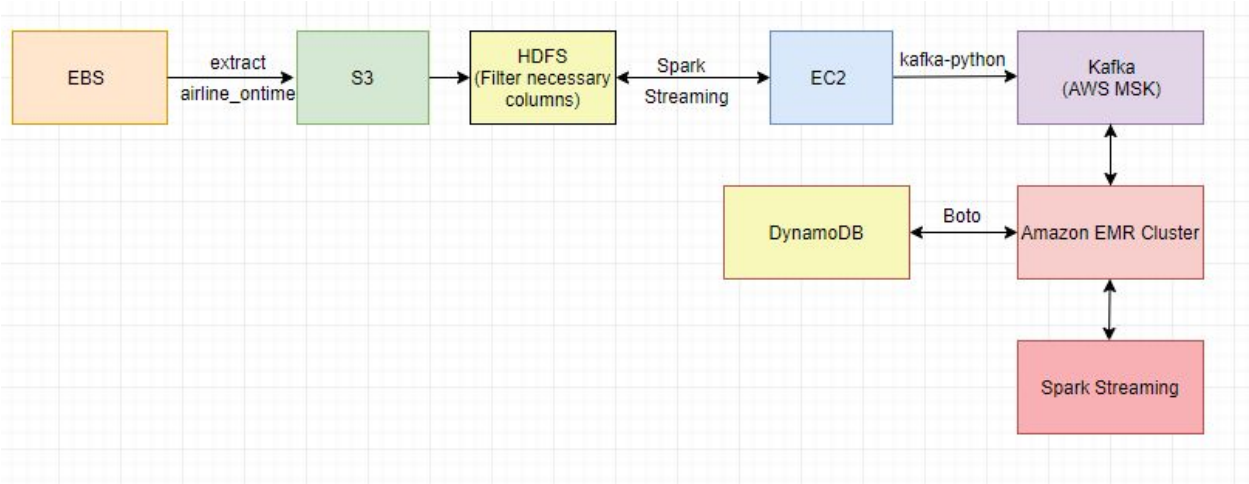
Overall Design



- For Task 2, I have adopted Kafka + Spark Streaming
- Airline_ontime folder is unzipped to S3.
- Relevant columns from CSV are filtered and stored in HDFS which is then streamed to EC2. EC2 instance then writes the data to Kafka topic
- AWS MSK is used to set up a Kafka Cluster with replication factor = 3 one m5.x2large node in each of the 3 regions. Kafka version used in 2.1.0
- Data in Kafka topic is consumed by EMR cluster, task2 queries are processed and the results stored in DynamoDB using boto python library.

Data Extraction

EBS Volume is mounted on EC2 and then only the required folder (airline_ontime) is unzipped to S3.

```
./ebs_to_s3.sh $EBS_MOUNTED_FOLDER $S3_BUCKET
```

Source
Extraction Script

Data in S3 is copied to HDFS and relevant columns filtered using spark streaming

Source
streamFiltering.py (discards cancelled flights)
streamFilteringWithCancelled.py (includes cancelled flights)
streamFiltering_2008.py (includes info only for 2008 flights and discards cancelled flights)

- Filtered data is moved to kafka cluster under different topics. Kafka-python library is used for this

  Some optimization flags used to prevent data drop and speed up write

  ```
  batch_size=98304 (default batch size is 16k)
  linger_ms=100
  acks='all' (acks set to false by default)
  ```

  Source
  kafkaProducer.py
  kafkaProducer_2008.py

- Kafka Consumer is EMR cluster with Spark 2.4.2 which reads from the topics and stores final answers in dynaoDB

Problem 1.1

Strategy:

1. map() each airport_id to (airport_id,1)
2. reduce() count for each airport_id to get (airport_id,count)
3. filter out top 10 airports with highest counts

```
spark-submit --executor-memory 6g --packages
org.apache.spark:spark-streaming-kafka-0-8_2.11:2.1.0 --conf
spark.streaming.backpressure.enabled=true --conf
spark.streaming.kafka.maxRatePerPartition=250000 g1_1.py
```

Source:
g1_1.py
Results:
Format: (AirportID, Count)

```
19/06/04 06:46:56 INFO DStreamGraph: Updated checkpoint data for time 1559628726000 ms
19/06/04 06:46:56 INFO CheckpointWriter: Submitted checkpoint of time 1559628726000 ms to writer queue
19/06/04 06:46:56 INFO CheckpointWriter: Saving checkpoint for time 1559628726000 ms to file 's3://mudabircapstonecheckpoint/top10airports/checkpoint-1559628726000'
(u'ORD', 12446097)
(u'ATL', 11537401)
(u'DFW', 10795494)
(u'LAX', 7721141)
(u'PHX', 6582467)
(u'DEN', 6270420)
(u'DTW', 5635421)
(u'IAH', 5478257)
(u'MSP', 5197649)
(u'SFO', 5168898)
19/06/04 06:46:57 INFO JobScheduler: Finished job streaming job 1559629098000 ms.0 from job set of time 1559629098000 ms
```

Problem 1.2

Strategy

1) Filter for all non cancelled flights and map (flightID, ArrivalDelay)
2) Get (flightID, (`delaySum,count,avgArrivalDelay`) //see updateStateByKey
3) Filter top 10 flights with best Arrival delay

Source:
g1_2.py

Results:

Format: FlightID [totalDepDelay, totalCount, AvgArrivalDelay]

(u'HA', (-264258.0, 261175, -1.01180434574519))
(u'AQ', (175282.0, 151507, 1.1569234424812056))
(u'PS', (60319.0, 41581, 1.4506385127822803))
(u'ML (1)', (328150.0, 69119, 4.747609195734892))
(u'PA (1)', (1570649.0, 293971, 5.342870555258852))
(u'F9', (1751640.0, 320468, 5.465881148819851))
(u'NW', (52626818.0, 9468157, 5.558295875321882))
(u'WN', (82237109.0, 14794127, 5.558767272986098))
(u'OO', (16842083.0, 2936047, 5.736312463662878))
(u'9E', (2713127.0, 462424, 5.8671846616957595))

```
19/06/08 09:37:05 INFO JobScheduler: Added jobs for time 1559985912000 ms
19/06/08 09:37:05 INFO JobScheduler: Starting job streaming job 1559985912000 ms.0 from job set of time 1559985912000 ms
(u'HA', (-264258.0, 261175, -1.01180434574519))
(u'AQ', (175282.0, 151507, 1.1569234424812056))
(u'PS', (60319.0, 41581, 1.4506385127822803))
(u'ML (1)', (328150.0, 69119, 4.747609195734892))
(u'PA (1)', (1570649.0, 293971, 5.342870555258852))
(u'F9', (1751640.0, 320468, 5.465881148819851))
(u'NW', (52626818.0, 9468157, 5.558295875321882))
(u'WN', (82237109.0, 14794127, 5.558767272986098))
(u'OO', (16842083.0, 2936047, 5.736312463662878))
(u'9E', (2713127.0, 462424, 5.8671846616957595))
19/06/08 09:37:05 INFO JobScheduler: Finished job streaming job 1559985912000 ms.0 from job set of time 1559985912000 ms
19/06/08 09:37:05 INFO JobScheduler: Total delay: 713.794 s for time 1559985912000 ms (execution: 0.132 s)
```

Problem 2.1

```
'''

        The incoming data format is

        Year|Month|date|DayofWeek|UniqueCarrier|FlightNum|Origin|Dest|CRSDeptime|DepD
        elay|ArrDelay
        '''
```

(For each origin airport X, rank top 10 carriers in decreasing order of ontime departure)
Strategy:

1) Filter for all non cancelled flights and map ((OriginAirport,carrier), DepDelay)
2) Get ((OriginAirpor,carrier), (depDelaySum,count, avgDepDelay)) // see `flightsDelay.updateStateByKey(updateFunction)`
3) For each OriginAirport, filter 10 best carriers in terms of Dep delay
4) Store results in dynamoDB
    - Main DB has OriginAirport as Partition key and Carrier as Sort key
    - Also created a Secondary index mapped to main DB with Partition key as OriginAirport and sort key as DepDelay. Hence querying the Secondary index returns results sorted in increasing order of DepDelay.

Source:
g2_1.py

Results:

Query: [Table] Top10CarriersTask2: Origin, Carrier

Origin	Carrier	DepDelay
SRQ	TZ	-0.381996974281
SRQ	XE	1.48976677772
SRQ	YV	3.40402193784
SRQ	AA	3.58326653307
SRQ	UA	3.95212206243
SRQ	US	3.96839828967
SRQ	TW	4.30467606502
SRQ	NW	4.85635924135
SRQ	DL	4.86917943416
SRQ	MQ	5.35058823529

Query: [Table] Top10CarriersTask2: Origin, Carrier

Origin	Carrier	DepDelay
CMH	DH	3.49111470113
CMH	AA	3.51526494895
CMH	NW	4.04155500526
CMH	ML (1)	4.36645962733
CMH	DL	4.71344133974
CMH	PI	5.20129487934
CMH	EA	5.93738938053
CMH	US	5.99220342136
CMH	TW	6.15909742531
CMH	YV	7.96119133574

Query: [Index] CarriersByDepDelayTask2: Origin, DepDelay

Origin	Carrier	DepDelay
JFK	UA	5.96832536487
JFK	XE	8.11373626374
JFK	CO	8.20120808165
JFK	DH	8.74298090807
JFK	AA	10.0978800018
JFK	B6	11.1270962227
JFK	PA (1)	11.5555935761
JFK	NW	11.6378177165
JFK	DL	11.98453575
JFK	TW	12.641537803

Query: [Index] CarriersByDepDelayTask2: Origin, DepDelay

Origin	Carrier	DepDelay
SEA	OO	2.70581965466
SEA	PS	4.72063933287
SEA	YV	5.12226277372
SEA	TZ	6.34500393391
SEA	US	6.41238418226
SEA	NW	6.49876240739
SEA	DL	6.5280229259
SEA	HA	6.8554526749
SEA	AA	6.93982738752
SEA	CO	7.09645886862

Query: [Index] CarriersByDepDelayTask2: Origin, DepDelay

Origin	Carrier	DepDelay
BOS	TZ	3.06379208506
BOS	PA (1)	4.44716479505
BOS	ML (1)	5.73477564103
BOS	EV	7.20813771518
BOS	NW	7.24518878651
BOS	DL	7.44120228111
BOS	XE	8.10292249047
BOS	US	8.68794683166
BOS	AA	8.72883304265
BOS	EA	8.90206833904

Problem 2.2

Strategy:

Same as problem 2.1 except replace carrier with DestAirport

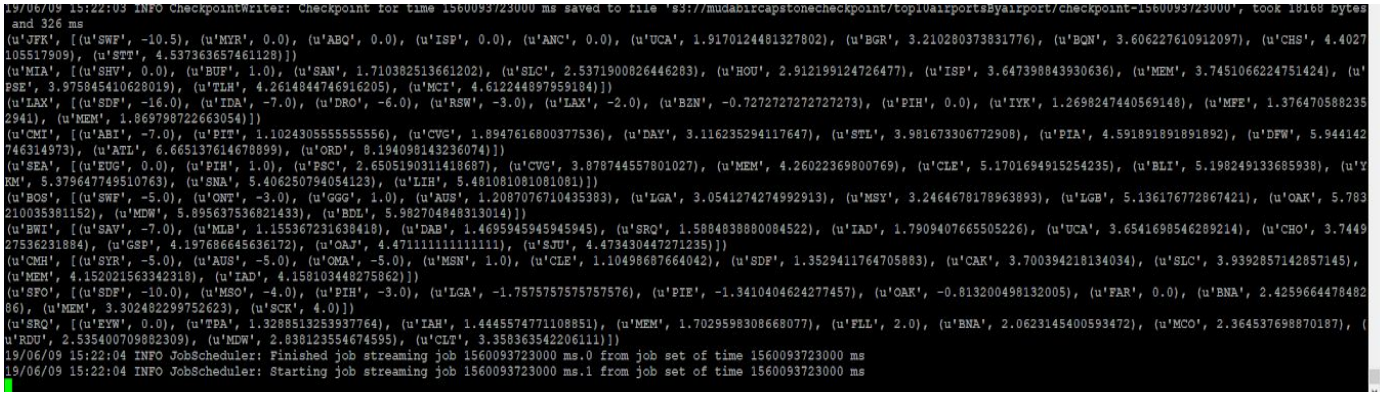
Source:

g2_2.py

Result:

Output format:

Origin, ((dest1, delay), (dest2,delay)....)



From logs:

and 322 ms

(u'JFK', [(u'SWF', -10.5), (u'MYR', 0.0), (u'ABQ', 0.0), (u'ISP', 0.0), (u'ANC', 0.0), (u'UCA', 1.9170124481327802), (u'BGR', 3.210280373831776), (u'BQN', 3.606227610912097), (u'CHS', 4.4027105517909), (u'STT', 4.537363657461128)])

(u'MIA', [(u'SHV', 0.0), (u'BUF', 1.0), (u'SAN', 1.710382513661202), (u'SLC', 2.5371900826446283), (u'HOU', 2.912199124726477), (u'ISP', 3.647398843930636), (u'MEM', 3.7451066224751424), (u'PSE', 3.975845410628019), (u'TLH', 4.2614844746916205), (u'MCI', 4.612244897959184)])

(u'LAX', [(u'SDF', -16.0), (u'IDA', -7.0), (u'DRO', -6.0), (u'RSW', -3.0), (u'LAX', -2.0), (u'BZN', -0.7272727272727273), (u'PIH', 0.0), (u'IYK', 1.2698247440569148), (u'MFE', 1.3764705882352941), (u'MEM', 1.869798722663054)])

(u'CMI', [(u'ABI', -7.0), (u'PIT', 1.1024305555555556), (u'CVG', 1.8947616800377536), (u'DAY', 3.116235294117647), (u'STL', 3.981673306772908), (u'PIA', 4.591891891891892), (u'DFW', 5.944142746314973), (u'ATL', 6.665137614678899), (u'ORD', 8.194098143236074)])

(u'SEA', [(u'EUG', 0.0), (u'PIH', 1.0), (u'PSC', 2.6505190311418687), (u'CVG', 3.878744557801027), (u'MEM', 4.26022369800769), (u'CLE', 5.1701694915254235), (u'BLI', 5.198249133685938), (u'YKM', 5.379647749510763), (u'SNA', 5.406250794054123), (u'LIH', 5.481081081081081)])

(u'BOS', [(u'SWF', -5.0), (u'ONT', -3.0), (u'GGG', 1.0), (u'AUS', 1.2087076710435383), (u'LGA', 3.0541274274992913), (u'MSY', 3.2464678178963893), (u'LGB', 5.136176772867421), (u'OAK', 5.783210035381152), (u'MDW', 5.895637536821433), (u'BDL', 5.982704848313014)])

(u'BWI', [(u'SAV', -7.0), (u'MLB', 1.155367231638418), (u'DAB', 1.4695945945945945), (u'SRQ', 1.5884838880084522), (u'IAD', 1.7909407665505226), (u'UCA', 3.6541698546289214), (u'CHO', 3.744927536231884), (u'GSP', 4.197686645636172), (u'OAJ', 4.471111111111111), (u'SJU', 4.473430447271235)])

(u'CMH', [(u'SYR', -5.0), (u'AUS', -5.0), (u'OMA', -5.0), (u'MSN', 1.0), (u'CLE', 1.10498687664042), (u'SDF', 1.3529411764705883), (u'CAK', 3.700394218134034), (u'SLC', 3.9392857142857145), (u'MEM', 4.152021563342318), (u'IAD', 4.158103448275862)])

(u'SFO', [(u'SDF', -10.0), (u'MSO', -4.0), (u'PIH', -3.0), (u'LGA', -1.7575757575757576), (u'PIE', -1.3410404624277457), (u'OAK', -0.8132004981320005), (u'FAR', 0.0), (u'BNA', 2.425966447848286), (u'MEM', 3.302482299752623), (u'SCK', 4.0)])

(u'SRQ', [(u'EYW', 0.0), (u'TPA', 1.3288513253937764), (u'IAH', 1.4445574771108851), (u'MEM', 1.7029598308668077), (u'FLL', 2.0), (u'BNA', 2.0623145400593472), (u'MCO', 2.364537698870187), (u'RDU', 2.535400709882309), (u'MDW', 2.838123554674595), (u'CLT', 3.358363542206111)])
19/06/09 15:21:24 INFO JobScheduler: Finished job streaming job 1560093684000 ms.0 from job set of time 1560093684000 ms
19/06/09 15:21:24 INFO JobScheduler: Starting job streaming job 1560093684000 ms.1 from job set of time 1560093684000 ms
19/06/09 15:21:25 INFO JobScheduler: Finished job streaming job 1560093684000 ms.1 from job set of time 1560093684000 ms
19/06/09 15:21:25 INFO JobScheduler: Total delay: 1.318 s for time 1560093684000 ms (execution: 1.206 s)
19/06/09 15:21:25 INFO KafkaRDD: Removing RDD 21220 from persistence list
19/06/09 15:21:25 INFO JobGenerator: Checkpointing graph for time 1560093684000 ms
19/06/09 15:21:25 INFO DStreamGraph: Updating checkpoint data for time 1560093684000 ms
19/06/09 15:21:25 INFO DStreamGraph: Updated checkpoint data for time 1560093684000 ms

Origin	Dest	DepDelay
JFK	SWF	-10.5
JFK	ISP	0
JFK	ABQ	0
JFK	MYR	0
JFK	ANC	0
JFK	UCA	1.91701244813
JFK	BGR	3.21028037383
JFK	BQN	3.60622761091
JFK	CHS	4.40271055179
JFK	STT	4.53736365746

Origin	Dest	DepDelay
SEA	EUG	0
SEA	PIH	1
SEA	PSC	2.65051903114
SEA	CVG	3.8787445578
SEA	MEM	4.26022369801
SEA	CLE	5.17016949153
SEA	BLI	5.19824913369
SEA	YKM	5.37964774951
SEA	SNA	5.40625079405
SEA	LIH	5.48108108108

Origin	Dest	DepDelay
CMH	SYR	-5
CMH	OMA	-5
CMH	AUS	-5
CMH	MSN	1
CMH	CLE	1.10498687664
CMH	SDF	1.35294117647
CMH	CAK	3.70039421813
CMH	SLC	3.93928571429
CMH	MEM	4.15202156334
CMH	IAD	4.15810344828

Origin	Dest	DepDelay
SRQ	EYW	0
SRQ	TPA	1.32885132539
SRQ	IAH	1.44455747711
SRQ	MEM	1.70295983087
SRQ	FLL	2
SRQ	BNA	2.06231454006
SRQ	MCO	2.36453769887
SRQ	RDU	2.53540070988
SRQ	MDW	2.83812355467
SRQ	CLT	3.35836354221

Origin	Dest	DepDelay
BOS	SWF	-5
BOS	ONT	-3
BOS	GGG	1
BOS	AUS	1.20870767104
BOS	LGA	3.0541274275
BOS	MSY	3.2464678179
BOS	LGB	5.13617677287
BOS	OAK	5.78321003538
BOS	MDW	5.89563753682
BOS	BDL	5.98270484831

Problem 2.4

(determine the mean arrival delay (in minutes) for a flight from X to Y)

Strategy:

1. Filter for all non cancelled flights and map (origin,dest) -> (ArrDelay)
2. Get (origin, dest) -> `(ArrDelaySum,count, avgArrDelay) //` See flightsDelay.updateStateByKey(updateFunction)
3. Save to DB

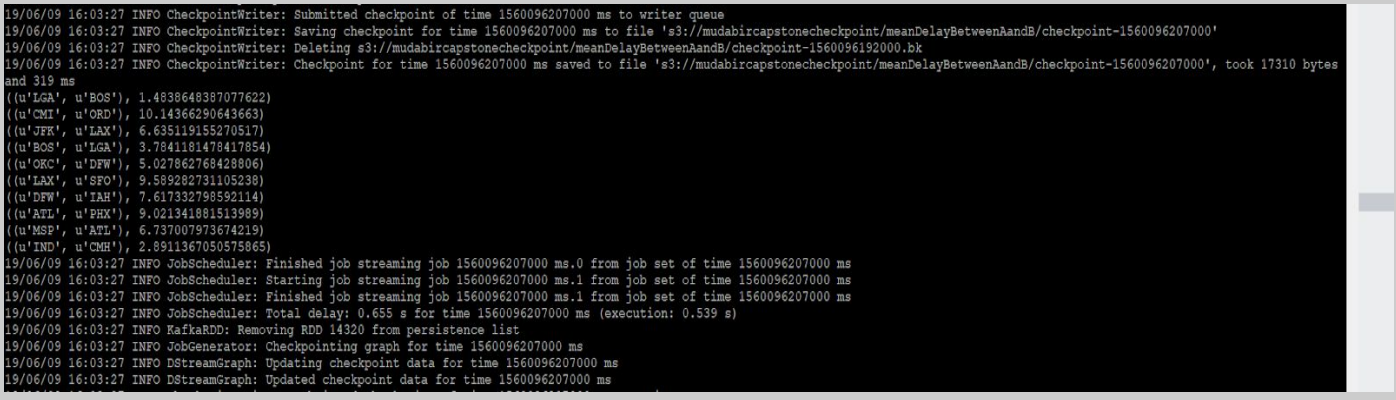
Source:

g2_4.py

Result:

Format: { (<origin>, <dest>) ,<delay value> }

((u'LGA', u'BOS'), 1.4838648387077622)
((u'CMI', u'ORD'), 10.14366290643663)
((u'JFK', u'LAX'), 6.635119155270517)
((u'BOS', u'LGA'), 3.7841181478417854)
((u'OKC', u'DFW'), 5.027862768428806)
((u'LAX', u'SFO'), 9.589282731105238)
((u'ATL', u'PHX'), 9.021341881513989)
((u'DFW', u'IAH'), 7.617332798592114)
((u'MSP', u'ATL'), 6.737007973674219)
((u'IND', u'CMH'), 2.8911367050575865)



Arrival delay of some origin-dest pairs from dynamoDB

Scan: [Table] MeanDelayBetweenAandBTask2: AtoB

AtoB	ArrDelay
(u'ATL', u'PHX')	9.02134188151
(u'BOS', u'LGA')	3.78411814784
(u'CMI', u'ORD')	10.1436629064
(u'DFW', u'IAH')	7.61733279859
(u'IND', u'CMH')	2.89113670506
(u'JFK', u'LAX')	6.63511915527
(u'LAX', u'SFO')	9.58928273111
(u'LGA', u'BOS')	1.48386483871
(u'MSP', u'ATL')	6.73700797367
(u'OKC', u'DFW')	5.02786276843

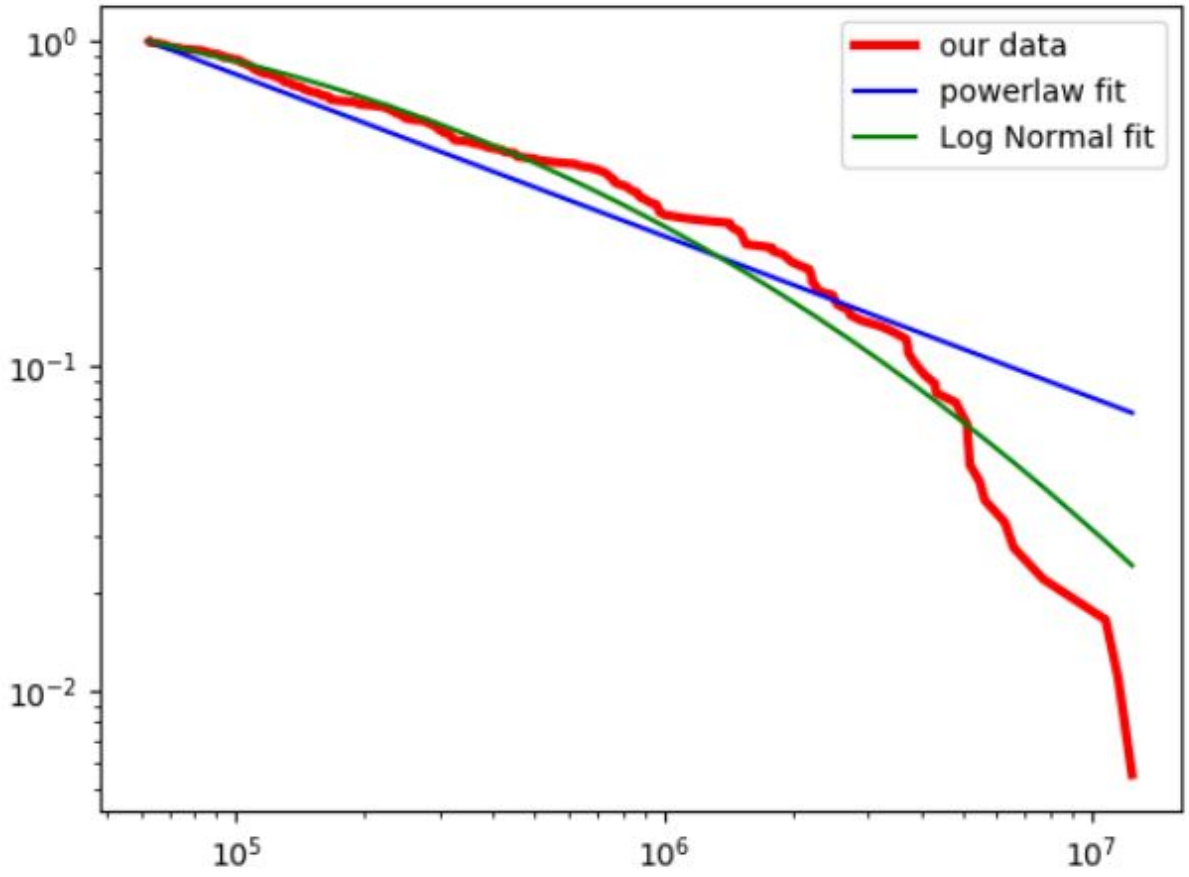
Problem 3.1
(copy pasting my results from Task1 for this problem)
Strategy:

1) Collected ranking of all airports into a file using Spark
2) Used powerlaw library to plot the CCDF of the power law distribution and the lognormal distribution of the data collected in 1)

Source:
g3_1.py
pythonplot.py

Result:
g3_1.log (Ranking of airports with their count)

Our data resembles the log Normal fit better than the powerlaw fit. We can conclude that the popularity of airports doesn't follow Zipf distribution



Problem 3.2

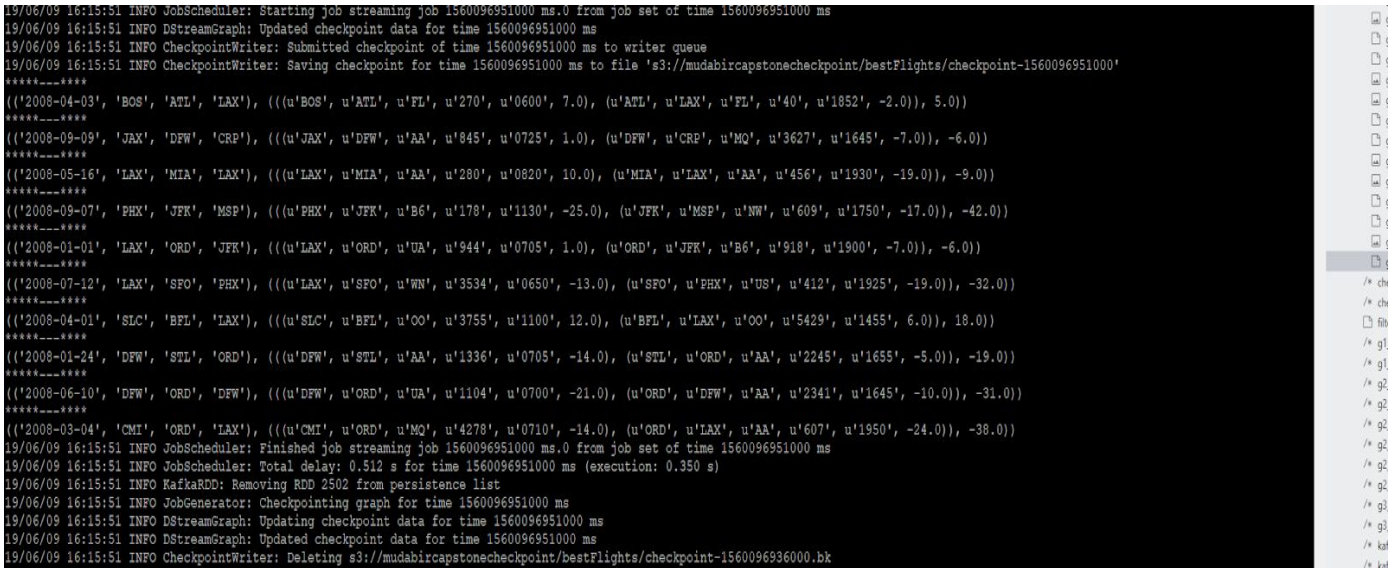
Strategy:

1. Filter non cancelled flights
2. Create RDD for all flights which fly before noon and dest as Y. key is (date, Y) -> (flight info)
3. Create RDD for all flights which fly after noon with origin as Y and departure date subtracted by 2 days
4. Join is done on (date,Y) as key. This gives all flights landing in Y before noon and all flights departing from Y two days later
5. For X->Y->Z route, filter the flight combo with minimum arrival delay
6. Write data to DynamoDB. Since a huge set of data was to be written, increased the write capacity of DynamoDB to 1500 to transfer the data faster.

Source:

g3_2.py

Result:



Format: ((startdate,X,Y,Z), ((XYdetails,YZdetails),totalArrivalDelay)

*****___****

(('2008-04-03', 'BOS', 'ATL', 'LAX'), (((u'BOS', u'ATL', u'FL', u'270', u'0600', 7.0), (u'ATL', u'LAX', u'FL', u'40', u'1852', -2.0)), 5.0))

*****___****

(('2008-09-09', 'JAX', 'DFW', 'CRP'), (((u'JAX', u'DFW', u'AA', u'845', u'0725', 1.0), (u'DFW', u'CRP', u'MQ', u'3627', u'1645', -7.0)), -6.0))

*****___****

(('2008-05-16', 'LAX', 'MIA', 'LAX'), (((u'LAX', u'MIA', u'AA', u'280', u'0820', 10.0), (u'MIA', u'LAX', u'AA', u'456', u'1930', -19.0)), -9.0))

*****___****

(('2008-09-07', 'PHX', 'JFK', 'MSP'), (((u'PHX', u'JFK', u'B6', u'178', u'1130', -25.0), (u'JFK', u'MSP', u'NW', u'609', u'1750', -17.0)), -42.0))

*****___****

(('2008-01-01', 'LAX', 'ORD', 'JFK'), (((u'LAX', u'ORD', u'UA', u'944', u'0705', 1.0), (u'ORD', u'JFK', u'B6', u'918', u'1900', -7.0)), -6.0))

*****___****

(('2008-07-12', 'LAX', 'SFO', 'PHX'), (((u'LAX', u'SFO', u'WN', u'3534', u'0650', -13.0), (u'SFO', u'PHX', u'US', u'412', u'1925', -19.0)), -32.0))

*****___****

(('2008-04-01', 'SLC', 'BFL', 'LAX'), (((u'SLC', u'BFL', u'OO', u'3755', u'1100', 12.0), (u'BFL', u'LAX', u'OO', u'5429', u'1455', 6.0)), 18.0))

*****___****

(('2008-01-24', 'DFW', 'STL', 'ORD'), (((u'DFW', u'STL', u'AA', u'1336', u'0705', -14.0), (u'STL', u'ORD', u'AA', u'2245', u'1655', -5.0)), -19.0))
*****___****
(('2008-06-10', 'DFW', 'ORD', 'DFW'), (((u'DFW', u'ORD', u'UA', u'1104', u'0700', -21.0), (u'ORD', u'DFW', u'AA', u'2341', u'1645', -10.0)), -31.0))
*****___****
(('2008-03-04', 'CMI', 'ORD', 'LAX'), (((u'CMI', u'ORD', u'MQ', u'4278', u'0710', -14.0), (u'ORD', u'LAX', u'AA', u'607', u'1950', -24.0)), -38.0))

Snapshot from DynamoDB

Scan: [Table] BestArrivalTimeFinalTask2: XYZ, StartDate — Viewing 1 to 10 items

XYZ	StartDate	ArrDelay	info
BOS-ATL-LAX	2008-04-03	5	((u'BOS', u'ATL', u'FL', u'270', u'0600', 7.0), (u'ATL', u'LAX', u'FL', u'40', u'1852', -2.0))
CMI-ORD-LAX	2008-03-04	-38	((u'CMI', u'ORD', u'MQ', u'4278', u'0710', -14.0), (u'ORD', u'LAX', u'AA', u'607', u'1950', -24.0))
DFW-ORD-DFW	2008-06-10	-31	((u'DFW', u'ORD', u'UA', u'1104', u'0700', -21.0), (u'ORD', u'DFW', u'AA', u'2341', u'1645', -10.0))
DFW-STL-ORD	2008-01-24	-19	((u'DFW', u'STL', u'AA', u'1336', u'0705', -14.0), (u'STL', u'ORD', u'AA', u'2245', u'1655', -5.0))
JAX-DFW-CRP	2008-09-09	-6	((u'JAX', u'DFW', u'AA', u'845', u'0725', 1.0), (u'DFW', u'CRP', u'MQ', u'3627', u'1645', -7.0))
LAX-MIA-LAX	2008-05-16	-9	((u'LAX', u'MIA', u'AA', u'280', u'0820', 10.0), (u'MIA', u'LAX', u'AA', u'456', u'1930', -19.0))
LAX-ORD-JFK	2008-01-01	-6	((u'LAX', u'ORD', u'UA', u'944', u'0705', 1.0), (u'ORD', u'JFK', u'B6', u'918', u'1900', -7.0))
LAX-SFO-PHX	2008-07-12	-32	((u'LAX', u'SFO', u'WN', u'3534', u'0650', -13.0), (u'SFO', u'PHX', u'US', u'412', u'1925', -19.0))
PHX-JFK-MSP	2008-09-07	-42	((u'PHX', u'JFK', u'B6', u'178', u'1130', -25.0), (u'JFK', u'MSP', u'NW', u'609', u'1750', -17.0))
SLC-BFL-LAX	2008-04-01	18	((u'SLC', u'BFL', u'OO', u'3755', u'1100', 12.0), (u'BFL', u'LAX', u'OO', u'5429', u'1455', 6.0))

Optimizations

- Filtered only necessary fields from all the csv reducing the total data size to 4.5GB from 15 GB
- In the kafka producer side, increased batch size and linger_ms to dispatch larger amounts of data at a time to the kafka cluster without waiting or ack. Also make ack= True to make sure no data is dropped/lost while writing to kafka topic
- Made replication factor =3 to make the kafka cluster reliable. Kept the number of partitions to 1. Increasing the number of partitions seemed to make the read from consumer side slower

Streaming versus Batch processing

- For the given data, batch processing seemed a better option. Since we had to compute top 10 aiports/carriers etc based on entire data, doing stream processing as the data arrives seems like an overkill.
- However, stream processing is useful in analysing live how the parameters are changing over time.

Comparing Stacks

Since I used Spark for task1 and Spark Streaming for task2, the stacks were more of less similar. Using hadoop Map-reduce for task 1 would have made the process very slow.
