Project Report

Data Analysis and Integration – 2022/2023

Group: 2

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Data Warehouse SQL

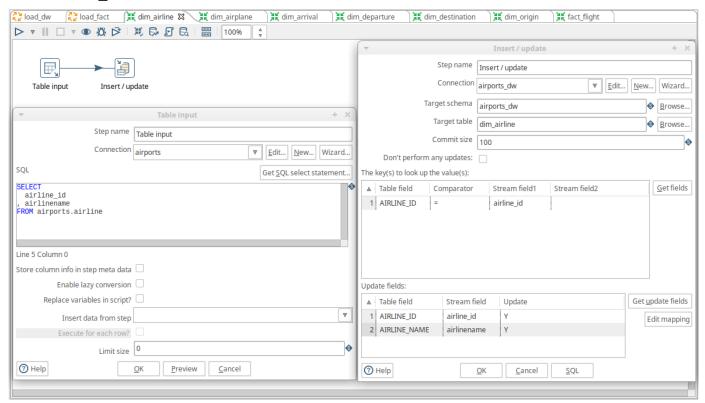
airports_dw.sql

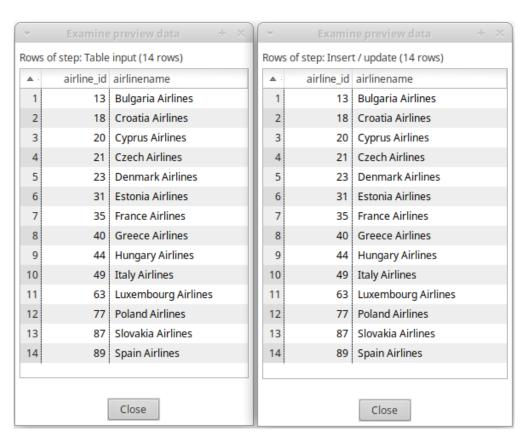
```
DROP DATABASE IF EXISTS airports_dw;
CREATE DATABASE airports_dw;
USE airports_dw;
CREATE TABLE dim_airport (
    AIRPORT_ID INT,
    AIRPORT_NAME VARCHAR(255),
    CITY VARCHAR(255),
    COUNTRY VARCHAR(255),
    PRIMARY KEY (AIRPORT_ID)
);
CREATE TABLE dim_departure (
    TIME_ID DATETIME,
    YEAR_ID INT,
    MONTH_ID INT,
    MONTH_NAME VARCHAR(255),
    DAY_ID INT,
    PRIMARY KEY (TIME_ID)
);
CREATE TABLE dim_arrival (
    TIME_ID DATETIME,
    YEAR_ID INT,
    MONTH_ID INT,
    MONTH_NAME VARCHAR(255),
    DAY_ID INT,
    PRIMARY KEY (TIME ID)
);
CREATE TABLE dim_airplane (
    AIRPLANE_ID INT,
    AIRPLANE_TYPE INT,
    PRIMARY KEY (AIRPLANE_ID)
);
CREATE TABLE dim_airline (
    AIRLINE ID INT,
    AIRLINE_NAME VARCHAR(255),
    PRIMARY KEY (AIRLINE_ID)
```

```
CREATE TABLE fact_flight (
    FLIGHT_ID INT,
    TOTALBOOKINGS INT,
    REVENUE DECIMAL(65,2),
    AIRLINE_ID INT,
    AIRPLANE_ID INT,
    ORIGIN_ID INT,
    DESTINATION_ID INT,
    DEPARTURE DATETIME,
    ARRIVAL DATETIME,
    PRIMARY KEY (FLIGHT_ID),
    FOREIGN KEY (AIRLINE_ID) REFERENCES dim_airline (AIRLINE_ID),
    FOREIGN KEY (AIRPLANE_ID) REFERENCES dim_airplane (AIRPLANE_ID),
    FOREIGN KEY (ORIGIN_ID) REFERENCES dim_airport (AIRPORT_ID),
    FOREIGN KEY (DESTINATION_ID) REFERENCES dim_airport (AIRPORT_ID),
    FOREIGN KEY (DEPARTURE) REFERENCES dim_departure (TIME_ID),
    FOREIGN KEY (ARRIVAL) REFERENCES dim_arrival (TIME_ID)
```

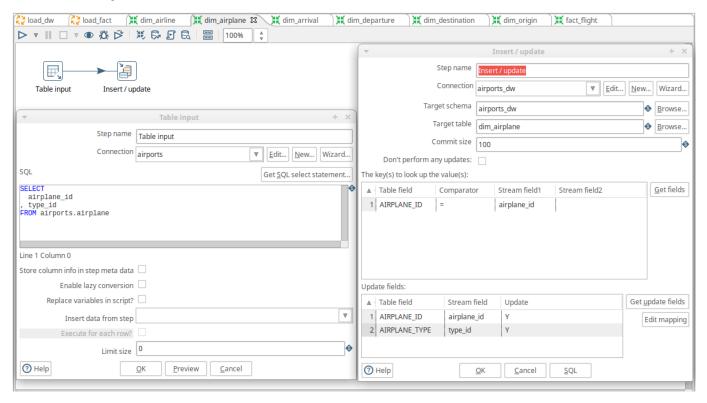
Transformations

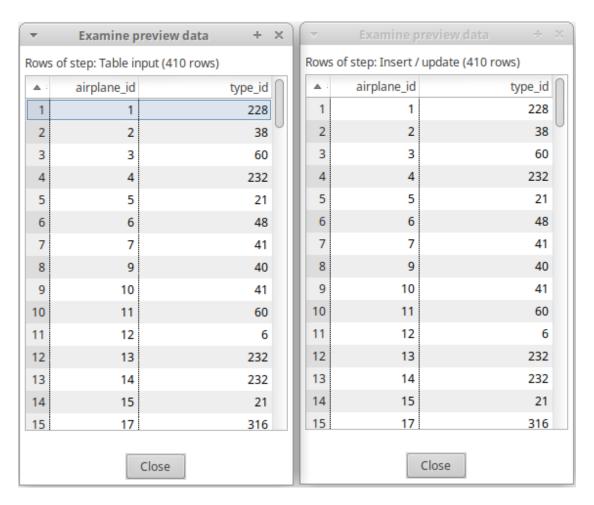
dim airline



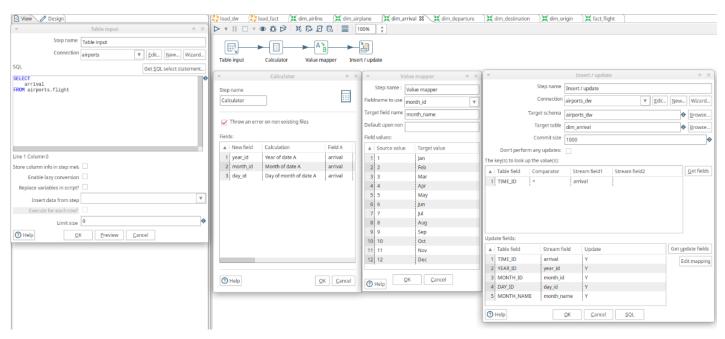


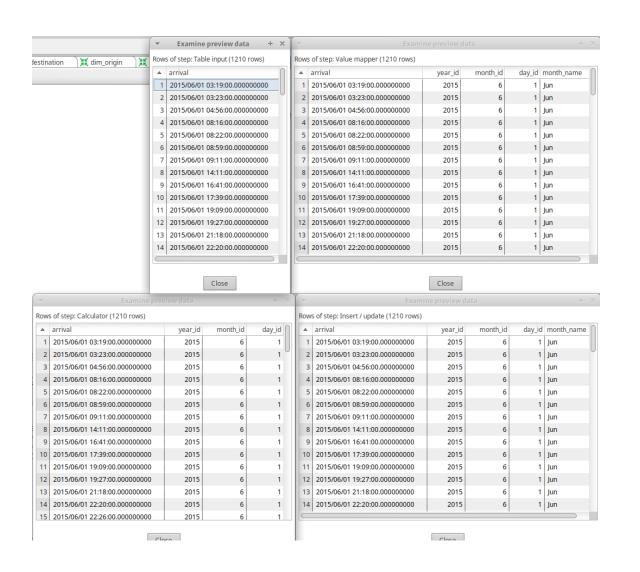
dim_airplane



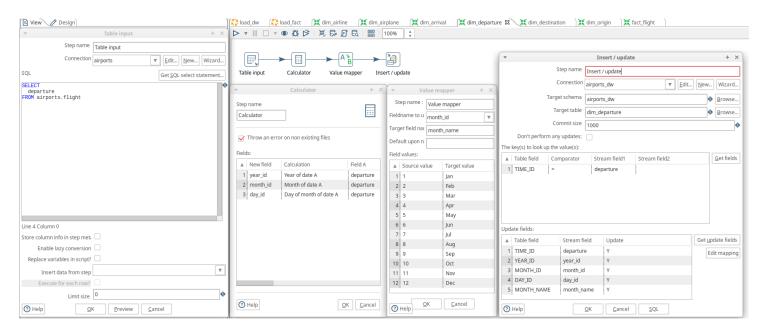


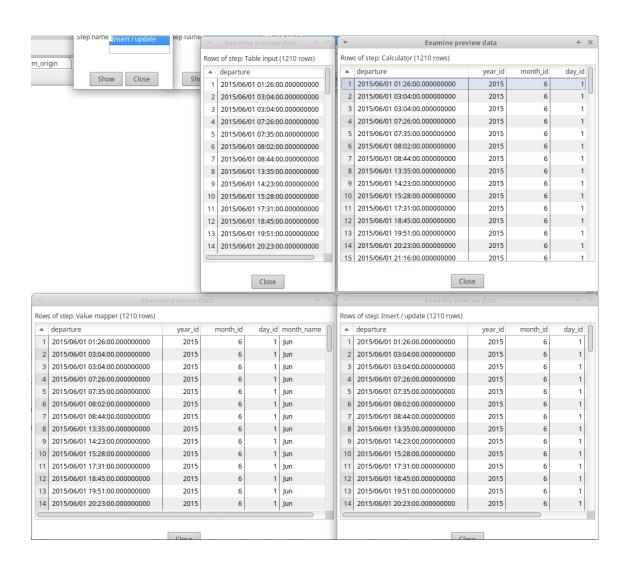
dim_arrival





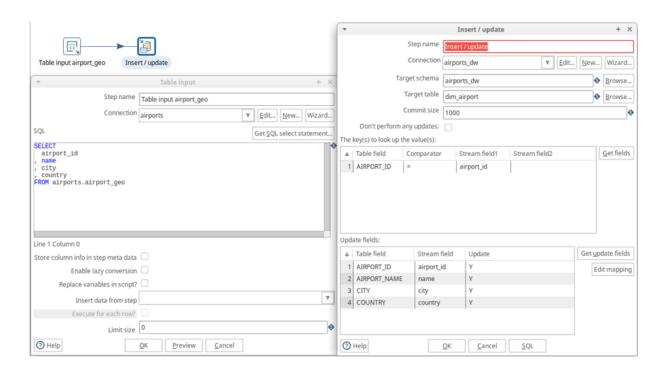
dim_departure

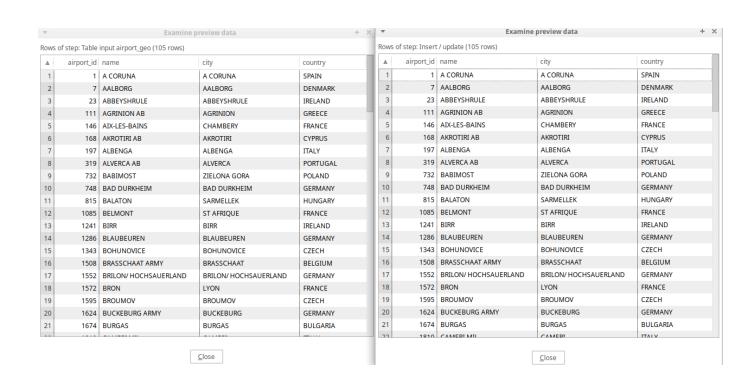




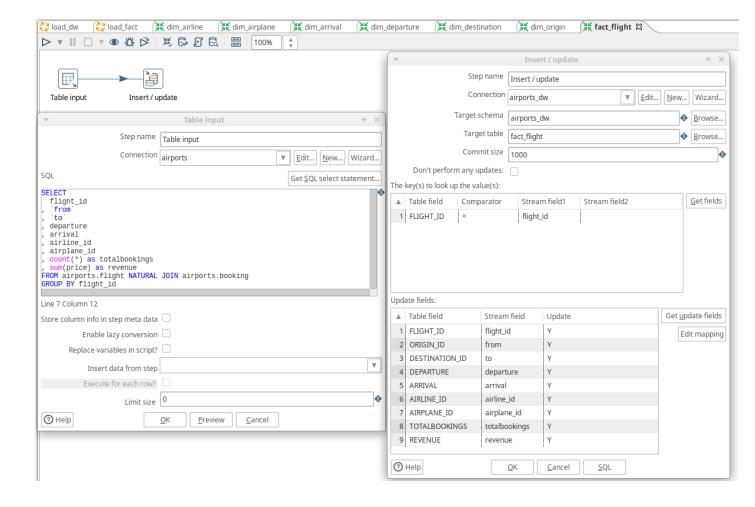
dim_airport

(Note: dim_origin and dim_destination were replaced by dim_airport.)





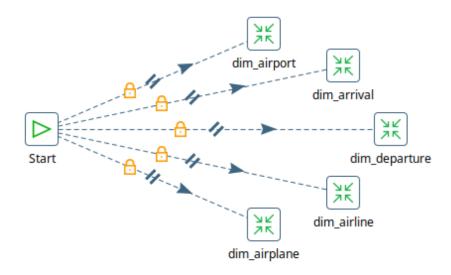
fact_flight



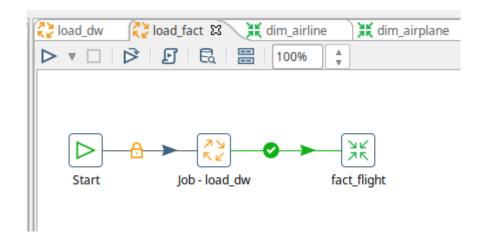
					Examine preview data				
WS (of step: Table	input (121	0 rows)						
. :	flight_id	from		departure	arrival	airline_id	airplane_id	totalbookings	revenue
1	750	12624	8266	2015/06/01 14:23:00.000000000	2015/06/01 16:41:00.000000000	18	3938	6	1757.93
2	899	4762	9633	2015/06/01 03:04:00.000000000	2015/06/01 04:56:00.000000000	21	4326	6	1794.15
3	1511	8591	1343	2015/06/01 18:45:00.000000000	2015/06/01 19:27:00.000000000	35	926	18	4166.23
4	1515	3420	748	2015/06/01 08:02:00.000000000	2015/06/01 08:22:00.000000000	35	902	4	651.13
5	1543	3887	10805	2015/06/01 20:23:00.000000000	2015/06/01 22:20:00.000000000	35	902	3	483.37
6	1871	1085	6973	2015/06/01 21:16:00.000000000	2015/06/01 22:26:00.000000000	44	2557	5	1544.67
7	1880	11812	6429	2015/06/01 01:26:00.000000000	2015/06/01 03:19:00.000000000	44	2564	17	3945.94
8	2117	1572	8829	2015/06/01 08:44:00.000000000	2015/06/01 09:11:00.000000000	49	1207	4	1347.69
9	2673	10564	6002	2015/06/01 22:50:00.000000000	2015/06/02 01:24:00.000000000	63	4869	21	4986.82
10	2703	4725	9838	2015/06/01 19:51:00.000000000	2015/06/01 21:18:00.000000000	63	4827	17	4490.16
11	2715	10904	2867	2015/06/01 17:31:00.000000000	2015/06/01 19:09:00.000000000	63	4840	6	1678.74
12	2717	8210	5024	2015/06/01 13:35:00.000000000	2015/06/01 14:11:00.000000000	63	4869	20	4245.1
13	3359	12159	3796	2015/06/01 15:28:00.000000000	2015/06/01 17:39:00.000000000	77	2255	18	4450.93
14	3367	10903	3074	2015/06/01 03:04:00.000000000	2015/06/01 03:23:00.000000000	77	2254	5	1469.83
					2015/06/01 00:50:00 000000000	0.7	4500	4	702.0
15	3793	2556	4430	2015/06/01 07:26:00.000000000	2015/06/01 08:59:00.000000000	87	4593	4:	/83.94 ÷
15 Rows	of step: Inser	rt / update ([1210 rows]		Close Examine preview data				÷
Rows	of step: Inser flight_id	rt / update (from	(1210 rows) to	departure	Close Examine preview data arrival	airline_id	airplane_id	totalbookings	revenue
Rows	of step: Inser flight_id 750	rt / update (from 12624	(1210 rows) to 8266	departure 2015/06/01 14:23:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000	airline_id	airplane_id 3938	totalbookings 6	revenue 1757.93
15 Rows	of step: Inser flight_id 750 899	rt / update (from 12624 4762	(1210 rows) to 8266 9633	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000	airline_id 18 21	airplane_id 3938 4326	totalbookings 6 6	revenue 1757.93 1794.15
15 Rows	of step: Inser flight_id 750 899 1511	rt / update (from 12624 4762 8591	(1210 rows) to 8266 9633 1343	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.000000000	airline_id 18 21 35	airplane_id 3938 4326 926	totalbookings 6 6 6 18	revenue 1757.93 1794.15 4166.23
15 Rows 1 2 3 4	of step: Inser flight_id 750 899 1511 1515	from 12624 4762 8591 3420	(1210 rows) to 8266 9633 1343 748	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.000000000 2015/06/01 08:22:00.000000000	airline_id 18 21 35 35	airplane_id 3938 4326 926 902	totalbookings 6 6 18 4	revenue 1757.93 1794.15 4166.23 651.13
15 Rows 1 2 3 4 5	of step: Inser flight_id 750 899 1511 1515 1543	rt / update (from 12624 4762 8591 3420 3887	(1210 rows) to 8266 9633 1343 748 10805	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 20:23:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.00000000 2015/06/01 08:22:00.000000000 2015/06/01 22:20:00.000000000	airline_id 18 21 35 35 35	airplane_id 3938 4326 926 902 902	totalbookings 6 6 18 4 3	revenue 1757.93 1794.15 4166.23 651.13 483.37
15 Rows 1 2 3 4 5 6	of step: Inser flight_id 750 899 1511 1515 1543 1871	rt / update (from 12624 4762 8591 3420 3887 1085	(1210 rows) to 8266 9633 1343 748 10805 6973	departure 2015/06/01 14:23:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 20:23:00.000000000 2015/06/01 21:16:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.00000000 2015/06/01 08:22:00.000000000 2015/06/01 22:20:00.000000000 2015/06/01 22:26:00.000000000	airline_id 18 21 35 35 35 44	airplane_id 3938 4326 926 902 902 2557	totalbookings 6 6 18 4 3 5	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67
Rows 1 2 3 4 5 6 7	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880	rt / update (from 12624 4762 8591 3420 3887 1085 11812	(1210 rows) to 8266 9633 1343 748 10805 6973 6429	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 20:23:00.000000000 2015/06/01 21:16:00.0000000000 2015/06/01 01:26:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.000000000 2015/06/01 08:22:00.000000000 2015/06/01 22:26:00.000000000 2015/06/01 22:26:00.000000000	airline_id 18 21 35 35 44 44	airplane_id 3938 4326 926 902 902 2557 2564	totalbookings 6 6 18 4 3 5	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94
15 Rows 1 2 3 4 5 6 7 8	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117	rt / update (from 12624 4762 8591 3420 3887 1085 11812 1572	(1210 rows) to 8266 9633 1343 748 10805 6973 6429 8829	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 20:23:00.000000000 2015/06/01 21:16:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 08:44:00.000000000	Close arrival 2015/06/01 16:41:00.00000000 2015/06/01 04:56:00.00000000 2015/06/01 08:22:00.000000000 2015/06/01 08:22:00.000000000 2015/06/01 22:20:00.000000000 2015/06/01 22:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 09:11:00.000000000	airline_id 18 21 35 35 44 44	airplane_id 3938 4326 926 902 902 2557 2564 1207	totalbookings 6 6 18 4 3 5 17	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69
15 Rows 1 2 3 4 5 6 7 8 9	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117 2673	rt / update (from 12624 4762 8591 3420 3887 1085 11812 1572 10564	(1210 rows) to 8266 9633 1343 748 10805 6973 6429 8829 6002	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 21:16:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 08:44:00.000000000 2015/06/01 22:50:00.0000000000000000000000000000000	Close arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.000000000 2015/06/01 08:22:00.000000000 2015/06/01 22:20:00.000000000 2015/06/01 22:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 09:11:00.0000000000 2015/06/02 01:24:00.0000000000	airline_id 18 21 35 35 44 44 49 63	airplane_id 3938 4326 926 902 902 2557 2564 1207 4869	totalbookings 6 6 18 4 3 5 17 4 21	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69 4986.82
15 Rows 1 2 3 4 5 6 7 8 9 10	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117 2673 2703	rt / update (from 12624 4762 8591 3420 3887 1085 11812 1572 10564 4725	(1210 rows) to 8266 9633 1343 748 10805 6973 6429 8829 6002 9838	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 21:16:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 08:44:00.000000000 2015/06/01 22:50:00.0000000000000000000000000000000	Close arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 19:27:00.000000000 2015/06/01 08:22:00.000000000 2015/06/01 22:20:00.000000000 2015/06/01 22:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 09:11:00.000000000 2015/06/02 01:24:00.0000000000 2015/06/01 21:18:00.0000000000	airline_id 18 21 35 35 35 44 44 49 63	airplane_id 3938 4326 926 902 902 2557 2564 1207 4869 4827	totalbookings 6 6 18 4 3 5 17 4 21	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69 4986.82 4490.16
Rows 1 2 3 4 5 6 7 8 9 10 11	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117 2673 2703	tt / update (from 12624 4762 8591 3420 3887 1085 11812 1572 10564 4725 10904	(1210 rows) to 8266 9633 1343 748 10805 6973 6429 8829 6002 9838 2867	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 20:23:00.000000000 2015/06/01 21:16:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 01:25:00.000000000 2015/06/01 19:51:00.000000000000000000000000000000000	Close arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 09:27:00.000000000 2015/06/01 08:22:00.000000000 2015/06/01 02:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 09:11:800.000000000 2015/06/01 12:18:00.000000000000000000000000000000000	airline_id 18 21 35 35 35 44 44 49 63 63	airplane_id 3938 4326 926 902 902 2557 2564 1207 4869 4827 4840	totalbookings 6 6 18 4 3 5 17 4 21 17 6	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69 4986.82 4490.16 1678.74
15 Rows 1 2 3 4 5 6 7 8 9 10 11 12	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117 2673 2703 2715 2717	rt / update (from 12624 4762 8591 3420 3887 1085 11812 1572 10564 4725 10904	(1210 rows) to 8266 9633 1343 748 10805 6973 6429 8829 6002 9838 2867 5024	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 20:23:00.00000000 2015/06/01 21:16:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 10:26:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 17:31:00.000000000 2015/06/01 13:35:00.000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 08:22:00.00000000 2015/06/01 02:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 02:24:00.000000000 2015/06/01 01:24:00.000000000 2015/06/01 01:24:00.000000000000000000000000000000000	airline_id 18 21 35 35 35 44 44 49 63 63 63	airplane_id 3938 4326 926 902 902 2557 2564 1207 4869 4827 4840 4869	totalbookings 6 6 18 4 3 5 17 4 21 17 6 20	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69 4986.82 4490.16 1678.74 4245.1
15 Rows 1 2 3 4 5 6 7 8 9 10 11 12 13	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117 2673 2703 2715 2717 3359	rt / update (from 12624 4762 8591 3420 3887 1085 11812 1572 10564 4725 10904 8210 12159	to 8266 9633 1343 748 10805 6973 6429 8829 6002 9838 2867 5024 3796	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 18:45:00.000000000 2015/06/01 20:23:00.000000000 2015/06/01 21:16:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 17:31:00.000000000 2015/06/01 15:28:00.000000000000000000000000000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 08:22:00.00000000 2015/06/01 08:22:00.000000000 2015/06/01 02:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 09:11:00.000000000 2015/06/01 21:18:00.000000000 2015/06/01 11:00.0000000000000000000000000000000	airline_id 18 21 35 35 35 44 44 49 63 63 63 63	airplane_id 3938 4326 926 902 902 2557 2564 1207 4869 4827 4840 4869 2255	totalbookings 6 6 18 4 3 5 17 4 21 17 6 20 18	revenue 1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69 4986.82 4490.16 1678.74 4245.1
15 Rows 1 2 3 4 5 6 7 8 9 10 11 12	of step: Inser flight_id 750 899 1511 1515 1543 1871 1880 2117 2673 2703 2715 2717	rt / update (from 12624 4762 8591 3420 3887 1085 11812 1572 10564 4725 10904	(1210 rows) to 8266 9633 1343 748 10805 6973 6429 8829 6002 9838 2867 5024	departure 2015/06/01 14:23:00.000000000 2015/06/01 03:04:00.000000000 2015/06/01 08:02:00.000000000 2015/06/01 20:23:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 01:26:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 19:51:00.000000000 2015/06/01 15:28:00.000000000 2015/06/01 15:28:00.000000000000000000000000000000000	Close Examine preview data arrival 2015/06/01 16:41:00.000000000 2015/06/01 04:56:00.000000000 2015/06/01 08:22:00.00000000 2015/06/01 02:26:00.000000000 2015/06/01 03:19:00.000000000 2015/06/01 02:24:00.000000000 2015/06/01 01:24:00.000000000 2015/06/01 01:24:00.000000000000000000000000000000000	airline_id 18 21 35 35 35 44 44 49 63 63 63	airplane_id 3938 4326 926 902 902 2557 2564 1207 4869 4827 4840 4869	totalbookings 6 6 18 4 3 5 17 4 21 17 6 20	1757.93 1794.15 4166.23 651.13 483.37 1544.67 3945.94 1347.69 4986.82 4490.16 1678.74 4245.1

Jobs

load_dw – Loads dimensions in parallel.



load_fact – Loads fact_flight table after all dimensions are loaded.



Cube Definition

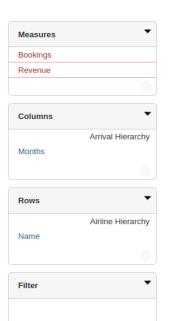
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  <Cube name="Flights" visible="true" cache="true" enabled="true">
    <Table name="fact flight">
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highCardinality="false" name="Origin">
      <Hierarchy name="Origin Hierarchy" visible="true" hasAll="true"</pre>
allMemberName="AllOrigins" primaryKey="AIRPORT ID">
        <Table name="dim airport">
        <Level name="Country" visible="true" column="COUNTRY" type="String"</pre>
uniqueMembers="false" levelType="Regular" hideMemberIf="Never">
        <Level name="City" visible="true" column="CITY" type="String"</pre>
uniqueMembers="false" levelType="Regular" hideMemberIf="Never">
        <Level name="Airport Name" visible="true" column="AIRPORT NAME" type="String"</pre>
uniqueMembers="false" levelType="Regular" hideMemberIf="Never">
        </Level>
      </Hierarchy>
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highCardinality="false" name="Destination">
      <Hierarchy name="Destination Hierarchy" visible="true" hasAll="true"</pre>
allMemberName="All Destinations" primaryKey="AIRPORT_ID">
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        </Table>
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uniqueMembers="false" levelType="Regular" hideMemberIf="Never">
        <Level name="City" visible="true" column="CITY" type="String"</pre>
uniqueMembers="false" levelType="Regular" hideMemberIf="Never">
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uniqueMembers="false" levelType="Regular" hideMemberIf="Never">
      </Hierarchy>
    <Dimension type="TimeDimension" visible="true" foreignKey="DEPARTURE"</pre>
highCardinality="false" name="Departure">
      <Hierarchy name="Departure Hierarchy" visible="true" hasAll="true"</pre>
allMemberName="All Departures" primaryKey="TIME_ID">
        <Table name="dim departure">
        </Table>
        <Level name="Year" visible="true" column="YEAR_ID" type="Integer"</pre>
uniqueMembers="false" levelType="TimeYears" hideMemberIf="Never">
```

```
<Level name="Months" visible="true" column="MONTH_NAME"</pre>
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hideMemberIf="Never">
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Saiku Analysis

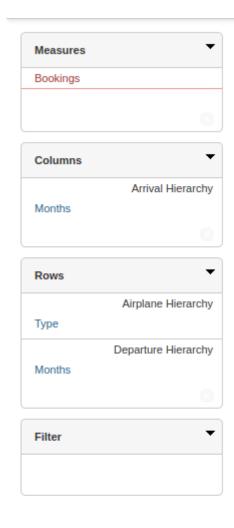
airports.sql

Passengers and revenue by airline and month.



Months		Jun		Jul	Aı	ug
Name	Bookings	Revenue	Bookings	Revenue	Bookings	Revenue
Bulgaria Airlines	449	\$ 110,113.24	587	\$ 154,437.99	-	-
Croatia Airlines	436	\$ 109,007.19	489	\$ 122,529.28	-	-
Cyprus Airlines	291	\$ 76,905.68	379	\$ 98,558.57	10	\$ 2,127.63
Czech Airlines	372	\$ 98,762.85	426	\$ 103,871.69	-	-
Denmark Airlines	187	\$ 46,222.46	236	\$ 55,558.29	-	-
Estonia Airlines	405	\$ 107,981.78	530	\$ 133,757.98	-	-
France Airlines	772	\$ 191,586.52	752	\$ 191,868.27	-	-
Greece Airlines	204	\$ 50,663.94	179	\$ 47,767.02	-	-
Hungary Airlines	1,089	\$ 274,893.57	1,143	\$ 286,874.33	-	-
Italy Airlines	165	\$ 42,343.58	210	\$ 56,486.71	-	-
Luxembourg Airlines	1,262	\$ 314,355.07	1,204	\$ 308,794.45	18	\$ 4,498.30
Poland Airlines	552	\$ 138,618.84	598	\$ 146,253.93	-	-
Slovakia Airlines	341	\$ 82,589.22	369	\$ 91,144.70	-	-
Spain Airlines	181	\$ 44,725.48	149	\$ 37,702.61	-	-

Airplane type and month of departure by month of arrival. (2 airplane types departed in July and arrived in August.)



	Months	Jun	Jul	Aug
Туре	Months	Bookings	Bookings	Bookings
6	Jun	1,354	-	-
	Jul	-	1,597	-
18	Jun	867	-	-
	Jul	-	931	-
21	Jun	885	-	-
	Jul	-	1,148	-
38	Jun	272	-	-
	Jul	-	219	10
40	Jun	129	-	-
	Jul	-	129	-
41	Jun	358	-	-
	Jul	-	343	-
48	Jun	328	-	-
	Jul	-	338	-
60	Jun	566	-	-
	Jul	-	731	18
75	Jun	135	-	-
	Jul	-	184	-
228	Jun	360	-	-
	Jul	-	361	-
232	Jun	129	-	-
	Jul	-	137	-
301	Jun	1,108	-	-
	Jul	-	976	-
316	Jun	215	-	-
	Jul	-	157	-

Money spent/gained on flights between countries. (On the 10 days with highest revenue – filter option.)

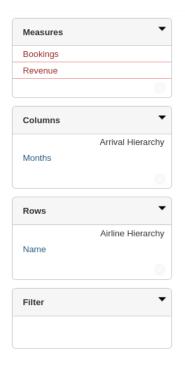
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Country	BELGIUM	CROATIA	CZECH	DENMARK	FINLAND	FRANCE	GERMANY	IRELAND	ITALY	POLAND	SPAIN	SWEDEN
Country	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue
AUSTRIA		-	-		-	-	-	-	-	\$ 27,681.59	-	
BELGIUM	-	-	-	-	-	-	\$ 7,676.85	-	-	-	-	-
CZECH	-	-	-	\$ 29,277.79	-	\$ 18,030.66	\$ 31,174.96	\$ 12,139.46	-	-	-	
DENMARK	-	-	-	-	-	-	-	-	\$ 5,285.41	-	-	-
ESTONIA	-	-	\$ 12,509.64	-	-	-	-	-	-	-	-	
FINLAND	-	-	-	-	-	-	\$ 45,667.59	-	-	-	-	
FRANCE	-	-	-	\$ 5,987.47	\$ 9,350.63	\$ 17,511.40	\$88,297.39	-	\$ 9,586.81	-	-	\$ 28,313.37
GERMANY	\$ 7,870.68	\$ 18,412.66	\$ 5,029.42	\$ 37,271.34	\$ 35,959.23	\$ 8,873.90	\$ 53,874.95	-	-	-	-	\$ 34,421.65
GREECE	-	-	-	-	-	\$ 4,880.61	-	-	\$ 12,256.94	-	-	
ITALY	-	-	-		-	\$ 9,363.88	\$ 12,908.67	-	-	-	\$ 5,490.45	-
POLAND	-	-	-	-	-	\$ 17,284.92	-	\$ 11,700.39	-	-	-	-
PORTUGAL	-	-	-	-	-	\$ 44,885.51	-	-	-	-	-	-
ROMANIA	-	-	-	-	\$ 12,032.21	-	-	-	-	\$ 6,261.86	-	
SPAIN	-	-	-	-	-	-	\$ 35,503.45	-	-	-	-	
SWEDEN	-	-			-	-	\$ 12,165.32			\$ 17,427.84	-	

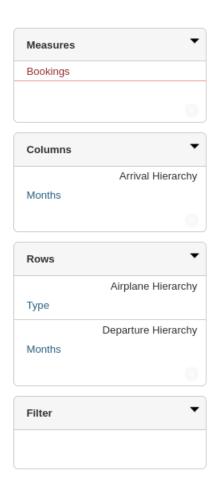
airports-large.sql

Passengers and revenue by airline and month.



Months		Jun		Jul	А	ug
Name	Bookings	Revenue	Bookings	Revenue	Bookings	Revenue
Afghanistan Airlines	15,352	\$ 3,864,611.80	16,118	\$ 4,062,950.72	318	\$ 81,725.14
Albania Airlines	15,951	\$ 3,991,510.15	17,053	\$ 4,308,360.83	229	\$ 57,998.03
American Samoa Airli	11,252	\$ 2,824,533.49	12,546	\$ 3,134,302.87	102	\$ 23,804.54
Angola Airlines	13,650	\$ 3,425,851.18	14,893	\$ 3,740,791.82	126	\$ 31,979.18
Argentina Airlines	15,661	\$ 3,947,144.14	17,098	\$ 4,309,724.79	274	\$ 67,685.16
Australia Airlines	14,726	\$ 3,686,163.92	15,853	\$3,987,009.91	271	\$ 65,439.72
Azerbaijan Airlines	11,636	\$ 2,892,437.23	12,449	\$ 3,138,111.32	206	\$ 49,668.41
Bahamas Airlines	16,613	\$ 4,171,537.98	17,918	\$ 4,513,922.77	135	\$ 35,741.87
Belarus Airlines	13,546	\$ 3,412,206.46	14,084	\$ 3,524,337.33	199	\$ 52,043.62
Bhutan Airlines	16,876	\$ 4,217,204.21	18,143	\$ 4,561,337.77	337	\$ 88,020.20
Bolivia Airlines	10,035	\$ 2,514,974.14	10,310	\$ 2,578,467.28	141	\$ 31,541.84
Brazil Airlines	18,264	\$ 4,590,913.75	18,679	\$ 4,698,229.33	268	\$ 65,752.29
Bulgaria Airlines	14,355	\$ 3,615,409.70	15,065	\$ 3,805,913.52	204	\$ 51,009.91
Caicos Is Airlines	13,899	\$ 3,504,475.25	13,677	\$ 3,391,208.05	176	\$ 43,858.96
Central African Rep	15,935	\$ 4,004,911.42	16,829	\$ 4,224,236.25	191	\$ 50,894.75
Chad Airlines	13,582	\$ 3,411,597.28	14,500	\$ 3,635,622.05	345	\$ 83,324.3
Colombia Airlines	13,396	\$ 3,362,594.33	14,495	\$ 3,665,361.21	157	\$ 40,559.92
Croatia Airlines	17,832	\$ 4,461,904.52	18,804	\$ 4,689,909.55	216	\$ 52,837.11
Cuba Airlines	14,201	\$ 3,553,389.12	14,849	\$ 3,716,627.79	89	\$ 24,691.91
Cyprus Airlines	18,091	\$ 4,534,422.14	19,375	\$ 4,875,355.50	306	\$ 78,392.56
Czech Airlines	14,426	\$ 3,613,421.07	15,658	\$ 3,951,107.66	173	\$ 44,100.96
Dakhla And Laayoune	14,152	\$ 3,560,507.07	14,659	\$ 3,676,618.99	278	\$ 68,369.9
Denmark Airlines	12,515	\$ 3,159,702.20	13,695	\$ 3,420,967.75	222	\$ 57,921.59
Djibouti Airlines	14,562	\$ 3,627,921.07	15,373	\$ 3,861,495.67	229	\$ 53,245.86
Dominica Airlines	11,570	\$ 2,900,001.11	11,594	\$ 2,909,360.74	114	\$ 28,687.2
Ecuador Airlines	15,418	\$ 3,886,397.82	16,016	\$ 4,029,561.24	240	\$ 59,310.6

Airplane type and month of departure by month of arrival. (2 airplane types departed in July and arrived in August.)



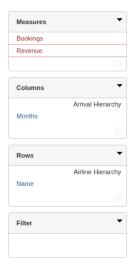
	Months	Jun	Jul	Aug
Туре	Months	Bookings	Bookings	Bookings
6	Jun	400,251	5,637	-
	Jul	-	403,462	5,273
18	Jun	235,757	3,324	-
	Jul	-	246,377	3,551
21	Jun	183,671	2,589	-
	Jul	-	189,362	2,534
	Aug	-	-	18
38	Jun	101,738	1,418	-
	Jul	-	106,392	1,701
	Aug	-	-	9
40	Jun	40,540	556	-
	Jul	-	41,693	512
41	Jun	58,249	770	-
	Jul	-	60,229	859
48	Jun	70,907	1,137	-
	Jul	-	73,456	968
60	Jun	220,202	3,451	-
	Jul	-	230,522	3,297
	Aug	-	-	18
75	Jun	50,143	700	-
	Jul	-	51,798	615
	Aug	-	-	11
228	Jun	78,211	963	-
	Jul	-	79,969	1,154
232	Jun	27,997	420	-
	Jul	-	28,914	427

Money spent/gained on flights between countries. (On the 10 days with highest revenue – filter option.)



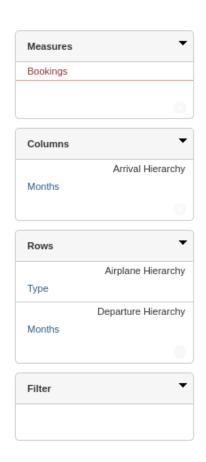
airports-large-extra.sql

Passengers and revenue by airline and month.



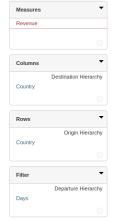
Months		Jun		Jul		Aug	Sep		
Name	Bookings	Revenue	Bookings	Revenue	Bookings	Revenue	Bookings	Revenue	
Afghanistan Airlines	153,531	\$ 38,484,790.79	161,293	\$ 40,472,054.67	162,391	\$ 40,865,554.24	8,124	\$ 2,039,445.5	
Albania Airlines	159,262	\$ 39,897,663.61	170,623	\$ 42,827,184.84	169,506	\$ 42,678,075.18	9,508	\$ 2,408,284.9	
American Samoa Airli	112,472	\$ 28,308,790.69	125,591	\$ 31,504,222.81	117,037	\$ 29,368,785.09	4,395	\$ 1,090,446.	
Angola Airlines	136,425	\$ 34,347,453.81	148,966	\$ 37,385,955.74	139,364	\$ 35,013,009.79	6,126	\$ 1,524,536.	
Argentina Airlines	156,706	\$ 39,433,100.48	171,009	\$ 42,933,031.41	158,448	\$ 39,820,229.90	7,249	\$ 1,792,335.	
Australia Airlines	147,264	\$ 36,895,382.74	158,391	\$ 39,717,352.17	154,653	\$ 38,775,426.65	7,486	\$ 1,881,800.	
Azerbaijan Airlines	116,404	\$ 29,158,202.24	124,498	\$ 31,339,897.35	124,711	\$ 31,326,210.56	6,458	\$ 1,613,125.	
Bahamas Airlines	166,189	\$ 41,667,568.68	179,311	\$ 45,011,524.23	170,440	\$ 42,797,620.62	8,441	\$ 2,124,317.	
Belarus Airlines	135,549	\$ 34,147,753.07	140,954	\$ 35,328,487.03	142,070	\$ 35,654,752.12	6,232	\$ 1,563,137	
Bhutan Airlines	168,839	\$ 42,317,956.17	181,527	\$ 45,582,515.26	175,595	\$ 44,040,959.30	10,125	\$ 2,546,693.	
Bolivia Airlines	100,221	\$ 25,079,321.56	103,110	\$ 25,822,166.50	103,653	\$ 26,018,204.52	5,146	\$ 1,296,724.	
Brazil Airlines	182,562	\$ 45,811,242.30	186,589	\$ 46,786,121.53	186,863	\$ 46,921,737.45	9,054	\$ 2,254,099	
Bulgaria Airlines	143,482	\$ 36,061,920.02	150,467	\$ 37,736,376.11	145,333	\$ 36,439,479.41	6,509	\$ 1,654,183	
Caicos Is Airlines	139,123	\$ 34,931,721.84	136,801	\$ 34,289,397.05	136,614	\$ 34,262,735.45	6,519	\$ 1,629,706	
Central African Rep	159,303	\$ 40,021,712.99	168,365	\$ 42,125,765.54	171,976	\$ 43,124,600.90	9,089	\$ 2,271,431	
Chad Airlines	135,782	\$ 34,072,100.08	144,879	\$ 36,371,560.39	145,014	\$ 36,362,031.39	6,079	\$ 1,536,691	
Colombia Airlines	133,934	\$ 33,595,180.27	144,968	\$ 36,335,348.56	144,710	\$ 36,238,243.43	7,400	\$ 1,878,148	
Croatia Airlines	178,186	\$ 44,646,913.61	188,029	\$ 47,132,982.06	187,251	\$ 46,927,131.22	9,023	\$ 2,296,912	
Cuba Airlines	142,058	\$ 35,617,149.64	148,646	\$ 37,354,343.65	148,108	\$ 37,196,850.77	6,600	\$ 1,651,337	
Cyprus Airlines	180,906	\$ 45,455,559.53	193,990	\$ 48,758,424.97	199,919	\$ 50,205,078.87	8,339	\$ 2,089,902	
Czech Airlines	144,527	\$ 36,249,272.89	156,292	\$ 39,263,017.41	150,664	\$ 37,883,903.19	7,735	\$ 1,936,701	
Dakhla And Laayoune	141,785	\$ 35,557,711.65	146,538	\$ 36,769,620.16	152,244	\$ 38,203,020.00	9,035	\$ 2,266,459.	
Denmark Airlines	125,033	\$31,357,361.07	136,844	\$ 34,271,543.51	130,576	\$ 32,805,019.70	5,613	\$ 1,412,421	
Djibouti Airlines	145,687	\$ 36,506,386.56	153,697	\$ 38,629,889.57	153,866	\$ 38,656,025.07	5,937	\$ 1,491,655.	
Dominica Airlines	115,913	\$ 29,190,379.84	115,808	\$ 29,054,151.34	120,969	\$ 30,433,939.52	5,332	\$ 1,332,572	
Ecuador Airlines	154,307	\$ 38,672,659.24	160,267	\$ 40,123,529.57	165,385	\$ 41,497,231.40	8,789	\$ 2,194,314	
Egypt Airlines	152,584	\$ 38,312,427.95	157,085	\$ 39,469,676.85	155,740	\$ 39,032,374.87	7,361	\$ 1,844,706	
El Salvador Airlines	152,013	\$ 38,186,942.96	157,311	\$ 39,541,539.00	162,757	\$ 40,862,191.03	6,817	\$ 1,713,911	
Equatorial Guinea Ai	175,440	\$ 44,094,510.06	178,659	\$ 44,785,869.08	185,993	\$ 46,608,700.52	9,920	\$ 2,472,782	
Eritrea Airlines	162,910	\$ 40,821,913.60	163,044	\$41,013,770.62	166,612	\$ 41,826,120.76	8,730	\$ 2,170,624	

Airplane type and month of departure by month of arrival. (2 airplane types departed in July and arrived in August.)



		_			_
	Months	Jun	Jul	Aug	Sep
Type	Months	Bookings	Bookings	Bookings	Bookings
6	Jun	4,002,690	56,319	-	-
	Jul	-	4,035,047	52,740	-
	Aug	-	-	4,103,815	59,872
	Sep	-	-	-	149,922
18	Jun	2,358,157	33,193	-	-
	Jul	-	2,463,421	35,468	-
	Aug	-	-	2,393,324	38,065
	Sep	-	-	-	81,851
21	Jun	1,837,798	25,875	-	-
	Jul	-	1,893,661	25,333	-
	Aug	-	-	1,917,024	27,036
	Sep	-	-	-	59,000
38	Jun	1,017,201	14,128	-	-
	Jul	-	1,064,500	16,928	-
	Aug	-	-	1,065,543	11,980
	Sep	-	-	-	36,378
40	Jun	405,278	5,555	-	-
	Jul	-	417,154	5,138	-
	Aug	-	-	420,771	6,087
	Sep	-	-	-	14,106
41	Jun	582,638	7,710	-	-
	Jul	-	602,124	8,583	-
	Aug	-	-	601,477	7,684
	Sep	-	-	-	19,412
48	Jun	709,642	11,393	-	-
	Jul	-	734,217	9,694	-
	Aug	-	-	728,225	9,743
	Sep	-	-	-	26,203
60	Jun	2,201,963	34,508	-	-
	Jul	-	2,304,932	32,936	-

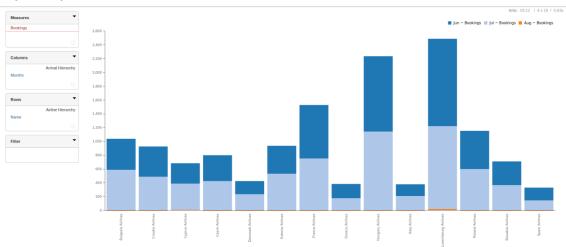
Money spent/gained on flights between countries. (On the 10 days with highest revenue – filter option.)



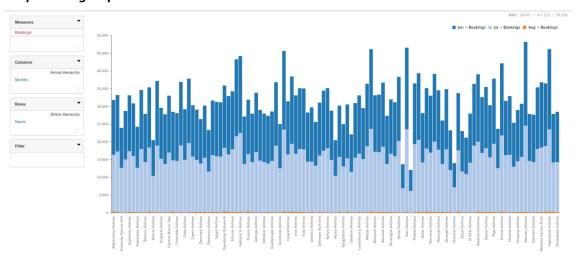
Country	AFGHANISTAN	ALGERIA	AMERICAN SAMOA	ANGOLA	ANGUILLA	ANTARCTICA (ARG)	ANTARCTICA (AUS)	ARGENTINA	ARMENIA	ARUBA	AUSTRALIA	AUSTRIA	AZORES
Country	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue	Revenue
AFGHANISTAN			-		-								
ALGERIA	-	-	-	-	-	-	-	\$ 386,102.32	-		-	-	-
AMERICAN SAMOA	-		-										
ANGOLA	-	-	-		-		-				\$ 261,284.63		
ANTARCTICA (ARG)	-	-	-	-	-	-	-					-	
INTARCTICA (AUS)	-		-										
ANTIGUA	-	-	-		-		-		-	-			-
ARGENTINA	-		-					\$ 331,643.03					
AUSTRALIA			-		-	-	-				\$1,796,940.11	\$1,004,368.41	
AUSTRIA	-	-	-	-	-		-				\$ 465,168.73		
AZERBAIJAN						-	-						
AZORES	-	-	-	-	-			-	-	-			
BAHAMAS			-										
BANGLADESH	-		-		-						\$ 537,843.54		
BELARUS	-	\$ 356,253.74	-		-								
BELGIUM											\$ 332,507.47		
BELIZE	-	-	-		-		-		-	-			-
BERMUDA						\$ 79,107.90							
BHUTAN	-		-		-								
BOLIVIA	\$ 15,098.57		-					\$ 261,156.05					
BOSNIA AND HERZEGOVINA						-	-						
BOTSWANA	-		-	-	-	-	-				\$ 246,207.10	\$ 239,058.34	
BRAZIL	\$ 706,268.94	\$ 69,465.56		\$ 361,301.95				\$ 961,243.61				\$ 375,928.95	\$ 314,347.42
RITISH VIRGIN IS	-	-	-	-	-								
BULGARIA	-				-								
BURKINA FASO													
BURUNDI	-	-	-	-	-					-			
CAICOS IS		\$ 47,779.15											

Comparisons

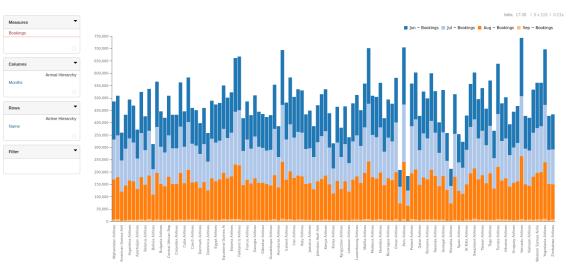
airports.sql



airports-large.sql



airports-large-extra.sql



We decided to compare the three datasets using the first query from exercise 5 (*Passengers and revenue by airline and month*), just with the bookings measure.

On an overall view, we can see the number of airlines growing considerably from the first to the second graphs, and then the appearance of a new month between the second and third ones.

Despite not comparing the number of airports and airplanes stored in the data warehouse, just by looking at the scales of the chosen graphs, we get a good idea of how the problem (manage information about flights) scales with the three .sql.

About the second query (*Airplane type and month of departure by month of arrival*), it only states the curiosity of the flights at the end of the month, which depart at one month and arrive at the other. The table only gets more lines as we increase the problem (can't be seen by the picture).

In the third query (*Money spent/gained on flights between countries*), we are able to see the scaling of big datasets, because between the first and the second, a considerable number of airports were introduced and the table became more sparse (a higher density of null values). At *airports-large-extra*, the number of airports barely changes (at least in the picture), but there are more flights between the same countries, as the table continues to be sparse and the money involved is higher.