



RESEARCH SCENARIO

Safety: Weather conditions can have a significant impact on aviation safety. Studying weather data can help airlines and aviation authorities predict and prepare for potential weather-related hazards such as turbulence, thunderstorms, or icing conditions.

Efficiency: Analyzing airplane data can help airlines and aviation authorities identify patterns and trends that can lead to improvements in fuel efficiency, flight times, and overall operational efficiency.

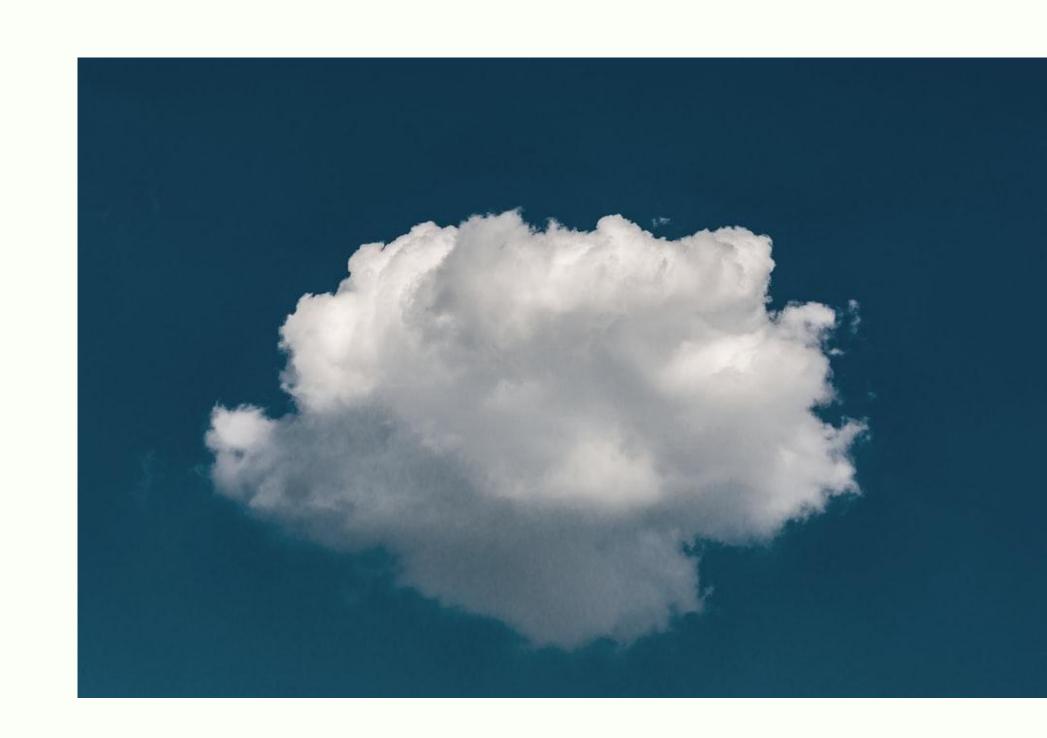
RESEARCH SCENARIO

Environmental impact: Researching weather and airplane data can also help to mitigate the environmental impact of aviation by reducing fuel consumption and emissions.

Improving technology: Studying weather and airplane data can help to identify areas for technological advancements and improvements in aircraft design and performance.

TOPICS

- 1. REQUIRMENTS
- 2. ETL
- 3. DATA QUALITY
- 4. ERD
- 5. SECURITY & COMPLIANCE
- 6. BACKUP & RECOVERY
- 7. TUNING & OPTIMIZATION
- 8. VISUAL & ANALYSIS
- 9. COMM THEORIES









2019 Flight Delays w/Weather and Airport

7	0 0800-0859	2	1	25	143 Southwest /	13056	107363	5873	1903352	13382999	6.18E-05	9.89E-05	8 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0700-0759	7	1	29	191 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	3 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	7	1	.27	199 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	18 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	9	1	.27	180 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	2 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0001-0559	7	1	10	182 Spirit Air L	13056	15023	1257	1903352	2688839	9.17E-06	0.00012465	1 McCarran I	36.08	-115.152 NONE	0	0	-0
7	0.0001-0559	.3	1	10	180 Frontier Air	13056	9496	581	1903352	1857122	0.00011573	7.13E-06	5 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0700-0759	6	1	29	186 Frontier Air	13056	9496	581	1903352	1857122	0.00011573	7.13E-06	2 McCarran I	36.08	-115.152 NONE	0	0	.0
7	1 0001-0559	7	1	10	186 Frontier Air	13056	9496	581	1903352	1857122	0.00011573	7.13E-06	3 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0001-0559	7	1	10	180 Frontier Air	13056	9496	581	1903352	1857122	0.00011573	7.13E-06	3 McCarran I	36.08	-115,152 NONE	0	0	-0
7	0 0600-0659	8	1	27	186 Frontier Ai	13056	9496	581	1903352	1857122	0.00011573	7.13E-06	1 McCarran I	36.08	-115.152 NONE	Ō	0	0
7	1 2300-2359	6	1	17	180 Frontier Ai	13056	9496	581	1903352	1857122	0.00011573	7.13E-06	3 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	1	1	27	149 Alaska Airl	13056	20315	717	1903352	2884187	3.23E-05	0.0001746	3 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 1200-1259	1	1	26	119 Alaska Airl	13056	20315	717	1903352	2884187	3.23E-05	0.0001746	12 McCarran I	36.08	-115.152 NONE	0	0	0
7	0.0600-0659	2	1	27	146 Alaska Airl	13056	20315	717	1903352	2884187	3.23E-05	0.0001746	7 McCarran I	36.08	-115.152 NONE	0	0	.0
7	0 0600-0659	4	1	.27	181 Alaska Airl	13056	20315	717	1903352	2884187	3.23E-05	0.0001746	2 McCarran I	36.08	-115.152 NONE	0	0	0
7	1 0700-0759	-4	1	29	181 Alaska Airl	13056	20315	717	1903352	2884187	3.23E-05	0.0001746	4 McCarran I	36.08	-115.152 NONE	-0	0	0
7	0 0800-0859	-4	1	25	181 Alaska Airl	13056	20315	717	1903352	2884187	3.23E-05	0.0001746	4 McCarran I	36.08	-115,152 NONE	0	0	-0
7	0 0001-0559	11	1	10	294 Hawaiian /	13056	6791	80	1903352	905990	0.00012049	0.00019785	6 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0900-0959	11	1	28	294 Hawaiian /	13056	6791	80	1903352	905990	0.00012049	0.00019785	5 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0700-0759	1	1	29	181 American A	13056	75506	1174	1903352	11744595	9.82E-05	0.00017729	5 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	8	1	27	187 American A	13056	75506	1174	1903352	11744595	9.82E-05	0.00017729	18 McCarran I	36.08	-115.152 NONE	0	0	-0
7	0 0800-0859	2	1	25	173 United Air I	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	6 McCarran I	36.08	-115.152 NONE	0	0	-0
7	0 0800-0859	3	1	25	142 United Air J	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	22 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0900-0959	7	1	28	154 United Air I	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	3 McCarran I	36.08	-115.152 NONE	0	0	0
7	1 1000-1059	3	1	29	142 United Air I	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	19 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	7	1	27	173 United Air I	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	4 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0700-0759	1	1	29	154 United Air J	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	21 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	2	1	.27	154 United Air J	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	20 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	9	1	27	173 United Air l	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	11 McCarran I	36.08	-115.152 NONE	0	0	-0
7	0 0001-0559	.5	1	10	154 United Air I	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	3 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	3	1	27	173 United Air I	13056	46218	1108	1903352	8501631	0.0002538	0.00022899	4 McCarran I	36.08	-115.152 NONE	0	0	.0
7	0 1000-1059	2	1	29	180 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	6 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	2	1	27	110 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	19 McCarran I	36.08	-115,152 NONE	0	0	-0
7	0 0700-0759	4	1	29	110 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	19 McCarran I	36.08	-115.152 NONE	Ö	0	0
7	0 0900-0959	6	1	.28	180 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0.00014866	1 McCarran I	36.08	-115.152 NONE	0	0	0
7	0 0600-0659	1	1	27	160 Delta Air L	13056	73508	1174	1903352	12460183	0.00014417	0,00014866	18 McCarran I	36.08	-115.152 NONE	0	0	0
7	1 2200-2259	9	1	9	162 JetBlue Airy	13056	23463	299	1903352	3190369	0.00016004	0,00012687	15 McCarran I	36.08	-115.152 NONE	0	0	0
7	0.0600-0659	9	1	27	129 JetBlue Airy	13056	23463	299	1903352			0.00012687	11 McCarran I	36.08	-115.152 NONE	0	0	0

A classification dataset with detailed airline, weather, airport and employment information.



Travel Dataset - Datathon 2019

avelCode	userCode	name	place	days	pric	e t	otal	dat	e							
-0		0 Hotel A	Florianopo		4	313.02	1252.08	09/	26/2019 hote	l do	ıta					
2		0 Hotel K	Salvador (B		2	263,41	526,82	10/	10/2019							
7	9	0 Hotel K	Salvador (B		.3	263,41	790.23	11/	14/2019							
11		0 Hotel K	Salvador (B		4	263,41	1053.64	12/	12/2019							
13		0 Hotel A	Florianopo		1	313.02	313.02	12/	26/2019							
15		0 Hotel BD	Natal (RN)		2	242.88	485.76	01/	09/2020							
22		0 Hotel Z	Aracaju (SE		2	208.04	416.08	02/	27/2020							
29		0 Hotel	n cons			binds	4054.00	21	(e.c. compo		_					_
32		0 Hotel tra	velCode userCo	-	locfrom	locto	flightT			_		listance	agency	date	270-2	
33		0 Hotel	-0	:0	Recife (PE) Florianc	po firstCk	155	1434.38		1.76	676.	53 FlyingDrop	09/26/2019	Fligh	data
34		0 Hotel	0	0	Florianope	Recife (I	PE) firstClc	iss	1292.29		1.76	676.	53 FlyingDrop	09/30/2019		
38		0 Hotel	1	_	Brasilia (D	+			1487.52		1.66		56 CloudFy	10/03/2019		
39		0 Hotel	1		Florianope		,		1127.36		1.66		56 CloudFy	10/04/2019		
			.2		Aracaju (S				1684.05		2.16		86 CloudFy	10/10/2019		
			2		Salvador (ISS	1531.92		2.16		86 CloudFy	10/12/2019		
			-3		Aracaju (S		cour	-	company	140	ame		gender	age		
			3		Campo Gr			0	4You	R	oy B	raun	male		21	user da
			4		Recife (PE			4	4V	1.	1	. II-1.			37	
			4		Florianope				4You			n Hols				
			5		Brasilia (D			2	4You	W	/ilm	a Meir	female		48	
			6	0 Aracaju (SE Bra 0 Recife (PE) Flor					4You	P	aula	Dani	female		23	
		-						4	4You	P	atric	ia Ca	female		44	
								5	4You	T	rina	Thon	none		47	
								6	4You	Je	sse I	Dece11	male		46	
								7	4You	G	irego	ria G	female		21	
									4You			abo			41	
									4You	-		ie Heli			35	
								-	4You			n Lov			36	
									4You				female		61	
											- 4					

A synthetic dataset of corporate travels



DATA AUDIT/USER STORY

The user table is mocked clients identification, it is not relevant to the research scenario and question. So it will not be upload and process in the following.

code	company	name	gender	age	
-0	4You	Roy Braun	male	21	user data
	4You	Joseph Hol	male	37	
2	4Yo.	Wilma Mci	female	48	
3	4You	Paula Dan:	female	23	
4	4You	Pan. 'a Ca	female	44	
- 5	4You	Trina Tho.	none	47	
-6	4You	Jesse Decel	mare	46	
7	4You	Gregoria G	female	21	
8	4You	Jack Sabo	none	41	
-9	4You	Debbie Hel	none	3.	
_10	4You	Melvin Lov	male	36	
11	4You	Virginia Ro	female	61	
12	4You	David Tho	male	53	







LOADING

```
COPY project.full_data_flightdelay (
MONTH,
DAY_OF_WEEK,
DEP_DEL15,
DEP_TIME_BLK,
DISTANCE_GROUP,
SEGMENT_NUMBER,
CONCURRENT_FLIGHTS,
NUMBER_OF_SEATS,
CARRIER_NAME,
AIRPORT_FLIGHTS_MONTH,
AIRLINE_FLIGHTS_MONTH,
AIRLINE_AIRPORT_FLIGHTS_MONTH,
AVG_MONTHLY_PASS_AIRPORT,
AVG_MONTHLY_PASS_AIRLINE,
FLT_ATTENDANTS_PER_PASS,
GROUND_SERV_PER_PASS,
PLANE_AGE,
DEPARTING_AIRPORT,
LATITUDE,
LONGITUDE,
PREVIOUS_AIRPORT,
PRCP,
SNOW,
SNWD,
TMAX,
AWND
FROM 'D:\Download\Spring2023\779\TERM PROJECT\full_data_flightdelay.csv'
DELIMITER ','
CSV HEADER;
```



CLEAN/TRANSFORM/FILTER

The data is clean, without anomalies, and well structured.

The data format is audited as follows:

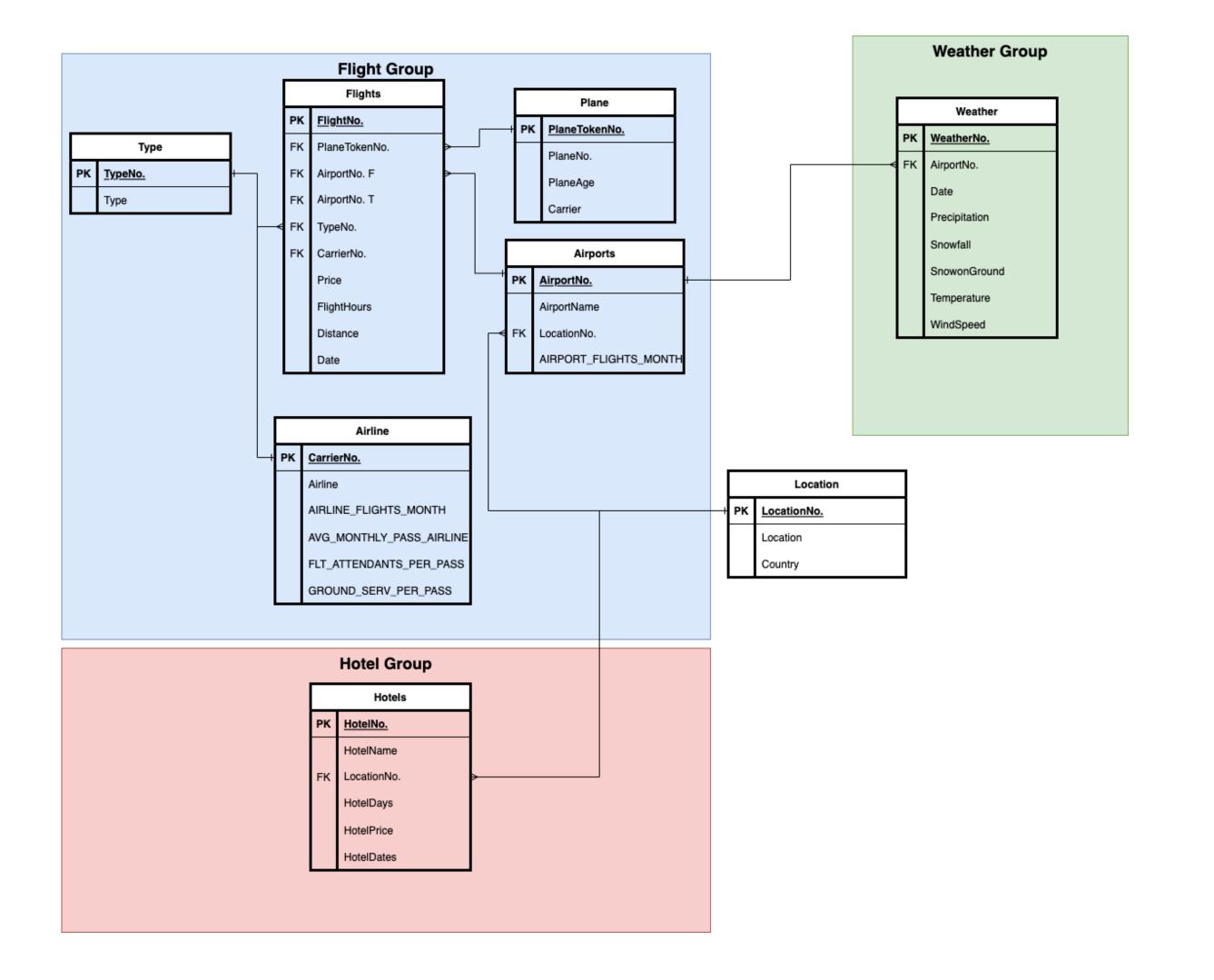
- all date DATE
- all time TIME
- flight number/weather report NUMERIC
- flight name/hotel name/name VARCHAR(25)
- PK SERIAL

<u>ALTER TABLE Project.flights</u> <u>ALTER COLUMN date TYPE date USING date::date,</u>

<u>ALTER TABLE Project.hotels</u> <u>ALTER COLUMN date TYPE date USING date::date;</u>







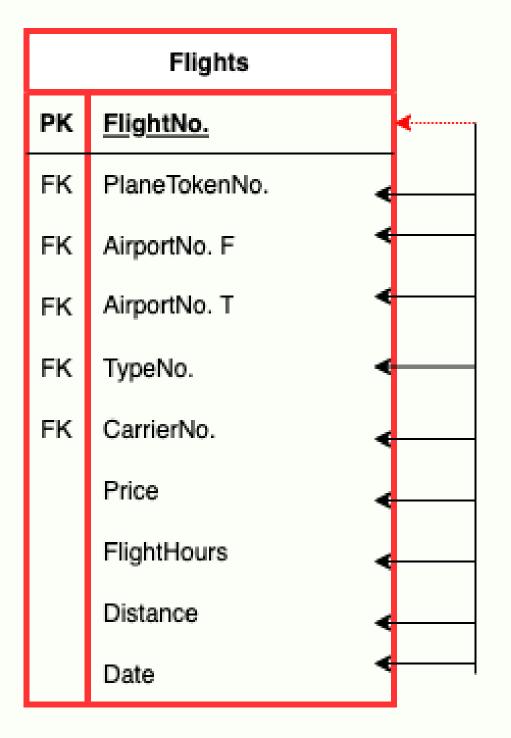
CREATING TABLES

```
CREATE TABLE project.type (
    TypeNo SERIAL PRIMARY KEY,
    TypePrevious varchar(25),
    TypeCurrent varchar(25)
);
```

INSERT INTO project.type (TypeCurrent)
SELECT DISTINCT project.f.flightType
FROM project.f



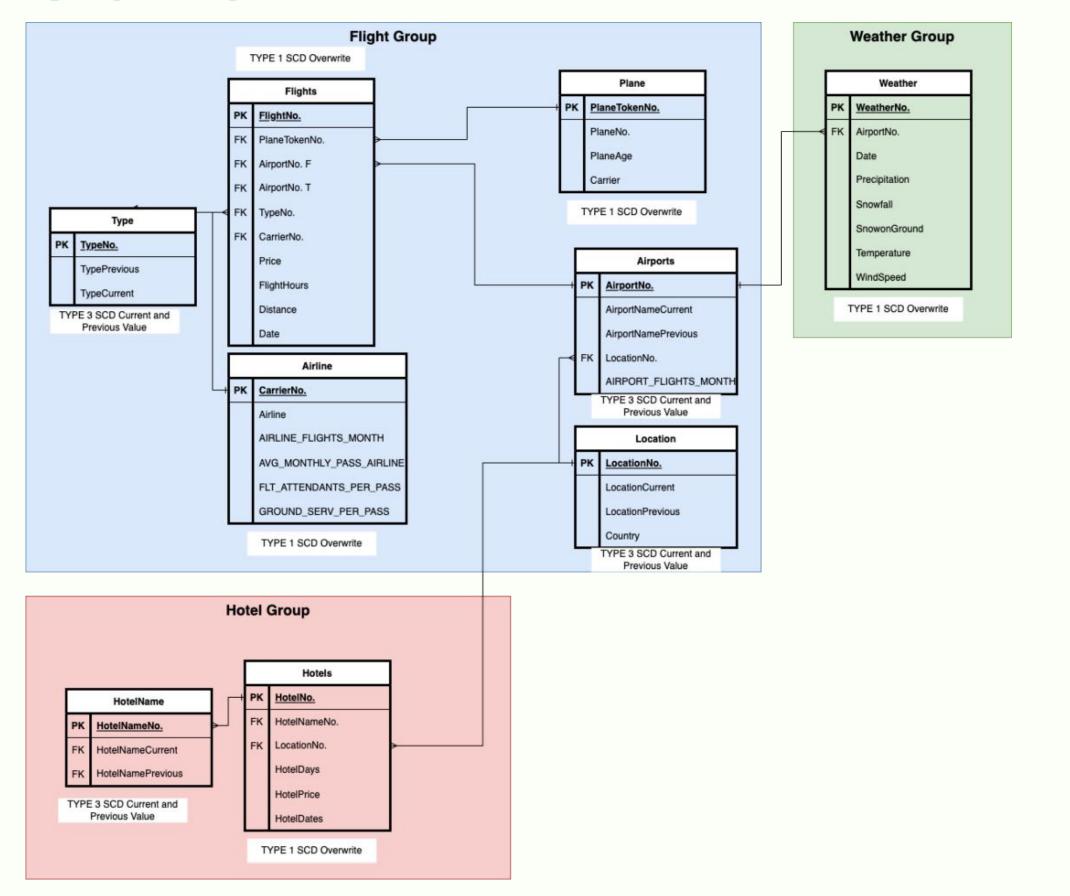
BCNF







SCDs







GDPR & HIPAA

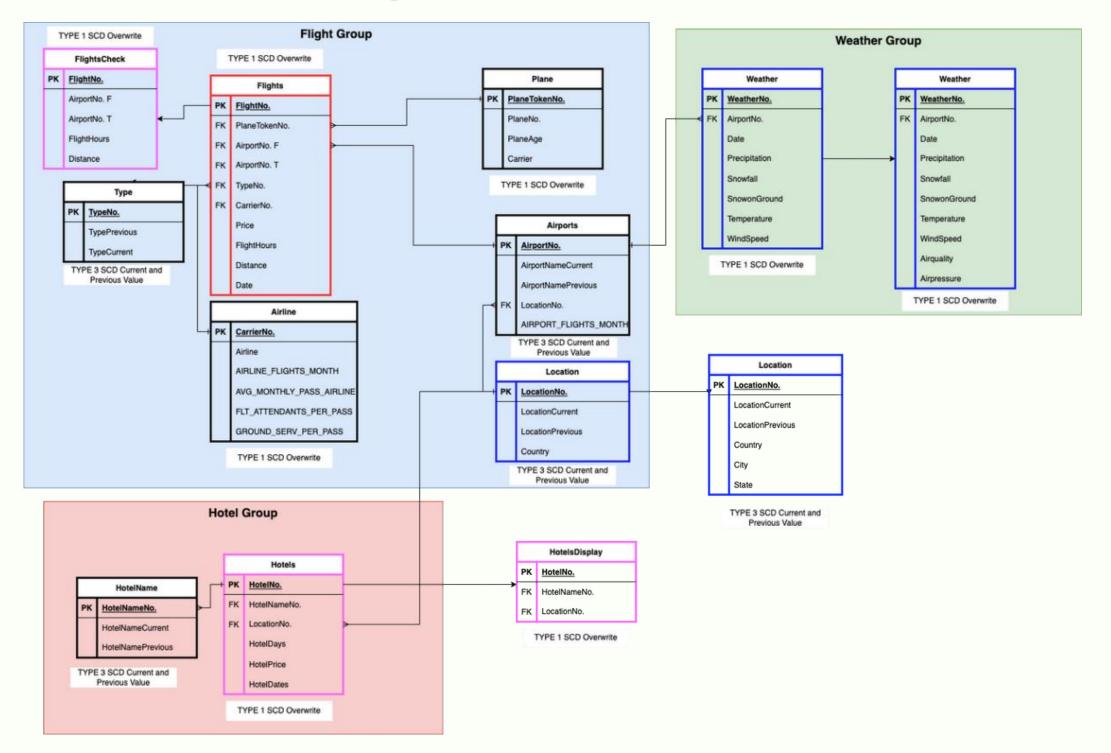
• The travel data does not have confidential data on medical records or personal privacy information.

• The dataset is GDPR & HIPAA-compliant.





PARTITION



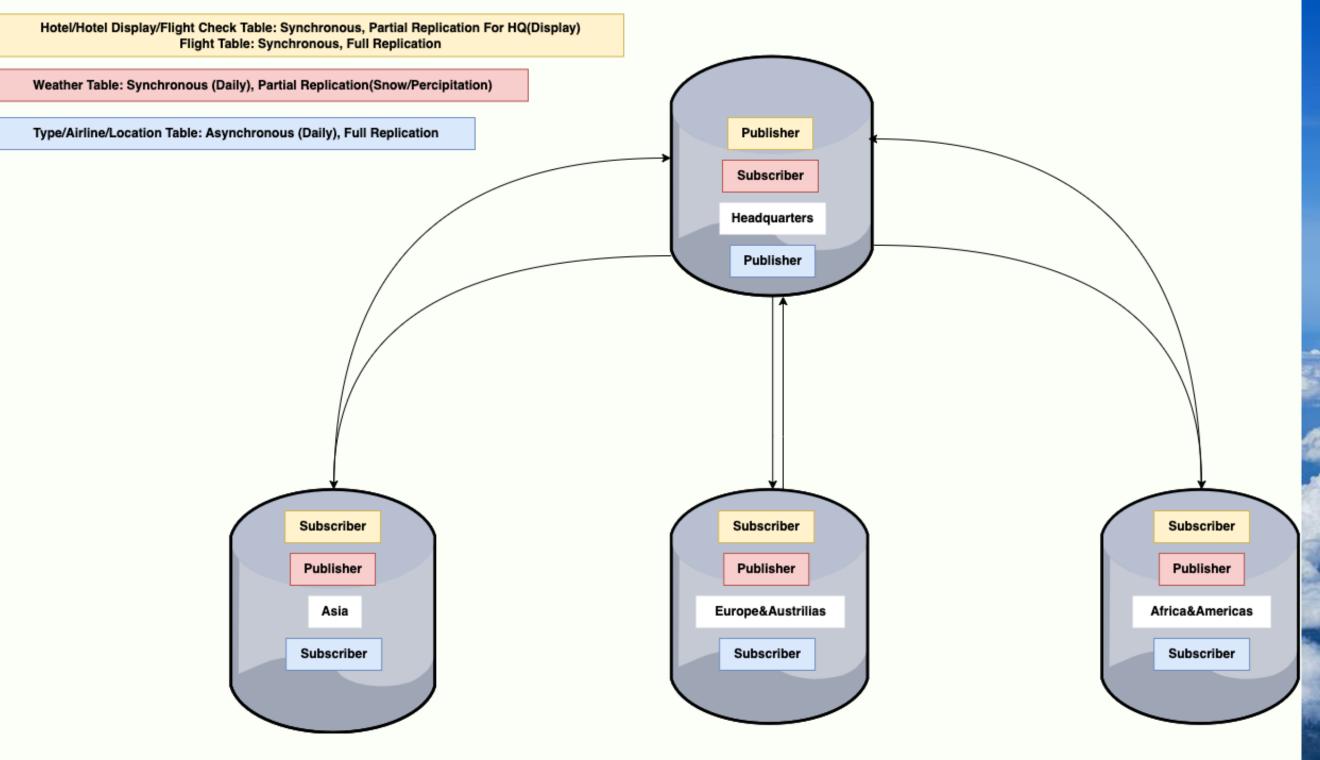
Red: Horizontal Partitioning (Shows local data for specific regions)

Blue: Vertical Partitioning(Modify attributes for regional local uses)

Pink: Vertical Partitioning(Show different tables for different department uses)

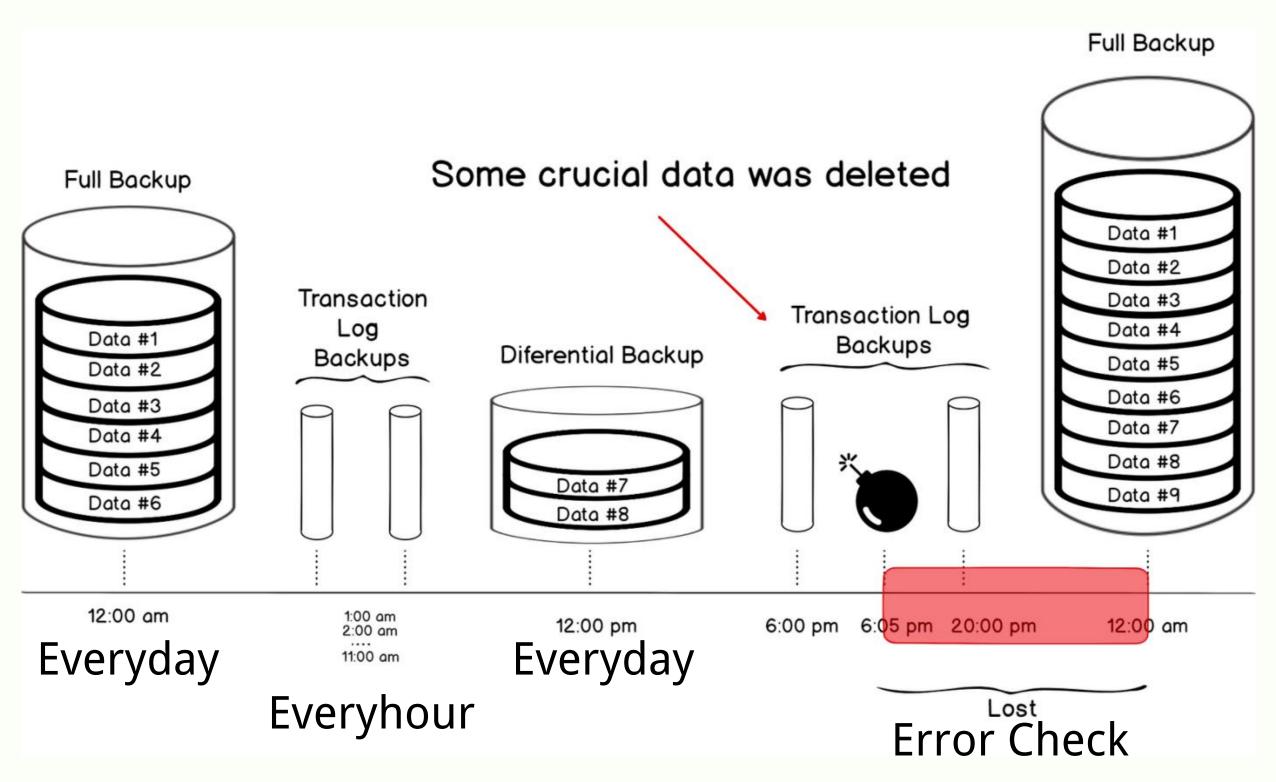


SYNCHRONIZATION





RECOVERY PLANS



https://www.sqlshack.com/sql-server-transaction-log-and-recovery-models/



DDT CHOICES

	Amazon Redshift	Microsoft Azure Synapse	Google BigQuery	Snowflake Cloud Data Platform
Initial Release	2012	2016	2010	2014
Separates Storage and Compute	No	Yes	Yes	Yes
Multi-Cloud	No	No	No	Yes
Query Language	Amazon Redshift SQL	TSQL	Standard SQL 2011 & BigQuery SQL	Snowflake SQL
Elasticity	Yes - Manual	Yes – Manual and Automatic	Yes – Automatic	Yes – Automatic
MPP	Yes	Yes	Yes	Yes
Columnar	Yes	Yes	Yes	Yes
Foreign Keys	Yes	Yes	No	Yes
Transaction	ACID	ACID	ACID	ACID
Concurrency	Yes	Yes	Yes	Yes
Durability	Yes	Yes	Yes	Yes
Automation	No	No	No	No
Website	Link	Link	Link	Link
Free Trial	Yes	Yes	Yes	Yes

https://www.sqlshack.com/sql-server-transaction-log-and-recovery-models/





HIGH AND LOW CARDINALITIES

- HIGH CARDINALITIES (a table has a large number of unique values)
- Flight
- Hotels
- Weather

- LOW CARDINALITIES (a table has)
- Airlines
- Planes



INDEXING-TYPE

- Data Types: Bitmap and B-tree indexes can handle both numeric and character data types, while hash indexes are typically limited to numeric data types
- Cardinality: Bitmap indexes are effective for low cardinality columns, while B-tree and hash indexes are suitable for high cardinality columns.
- Size and Storage: Bitmap indexes require less storage space compared to B-tree and hash indexes. However, B-tree indexes are more flexible and can handle large data sets. Hash indexes are typically faster than B-tree and bitmap indexes for exact match queries, but they are less flexible and less efficient for range queries.
- Query Performance: Bitmap indexes are useful for boolean operations such as AND, OR, and NOT, while B-tree and hash indexes are better suited for range queries and exact match lookups, respectively.
- Maintenance: B-tree and hash indexes are easier to maintain and update compared to bitmap indexes, which can become complex to maintain as data sets grow in size.

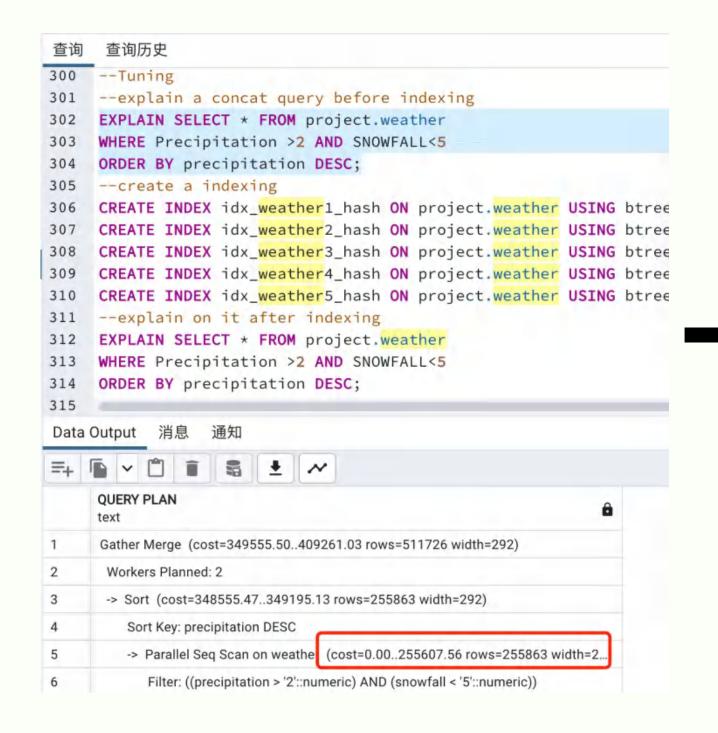


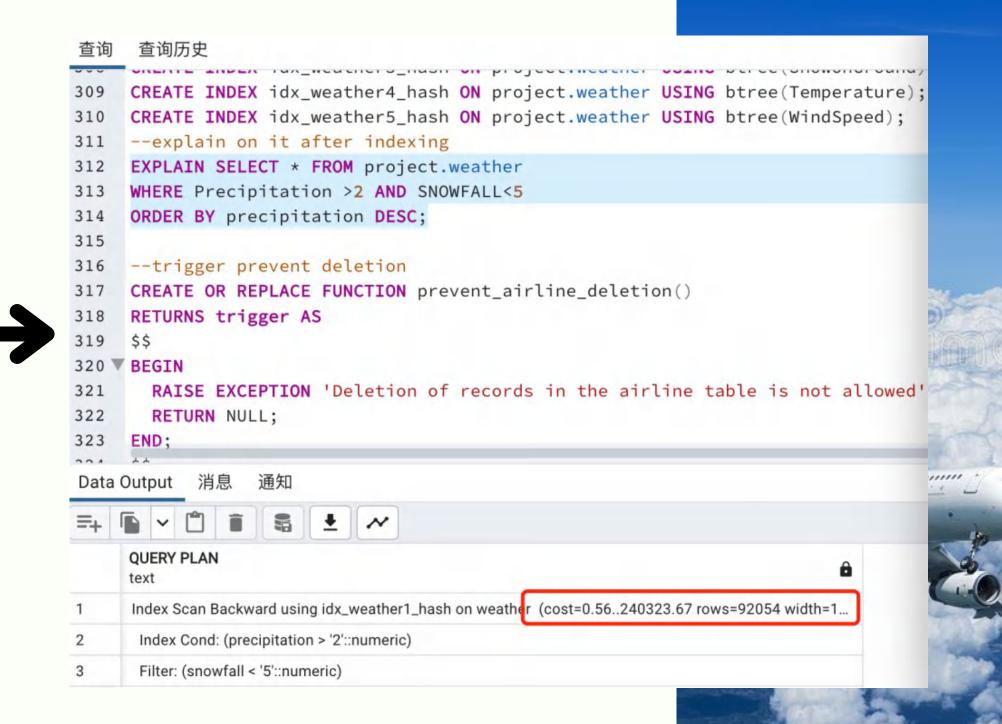
INDEXING-IMPLEMENTATION

- CREATE INDEX idx weather2 hash ON project.weather
 USING btree(Snowfall);
- CREATE INDEX idx weather3 hash ON project.weather
 USING btree(SnowonGround);
- <u>CREATE INDEX idx weather4 hash ON project.weather</u> <u>USING btree(Temperature)</u>;
- <u>CREATE INDEX idx weather5 hash ON project.weather</u>
 <u>USING btree(WindSpeed)</u>



INDEXING-IMPLEMENTATION



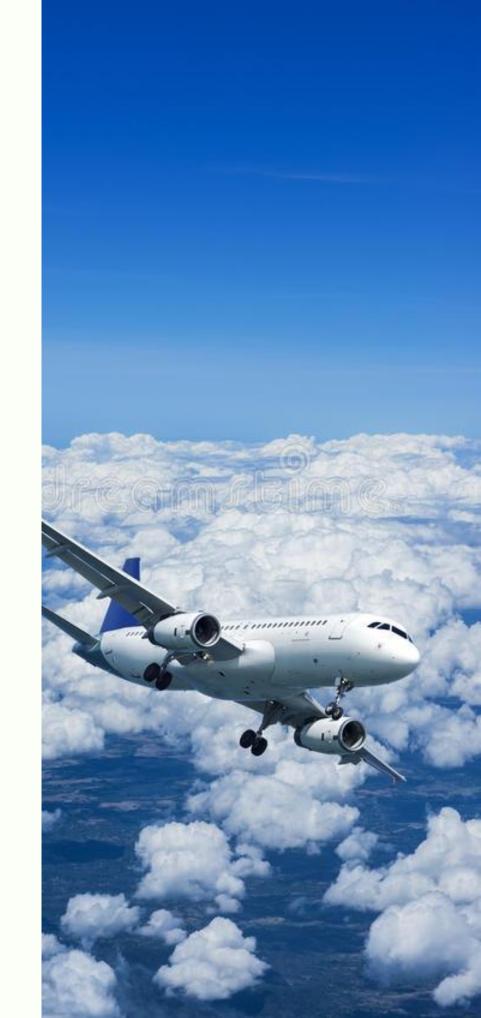


INDEXING-IMPROVEMENTS

• Index Combination: In some cases, creating a combined index on multiple columns can be more effective than creating separate indexes on each individual column. This can be particularly useful if queries often filter on multiple columns. For example, you might create an index on instead of separate indexes on and .

(Temperature, Precipitation)TemperaturePrecipitation

• Index Type Selection: As mentioned before, the hash index may be more efficient for columns with a small number of distinct values. For example, if has only a few distinct values, it may be more efficient to create a hash index on this column. Airquality



INDEXING-IMPROVEMENTS

- Index Maintenance: Regularly maintaining the indexes can help to ensure their efficiency. This can include reindexing the table to eliminate fragmentation, or updating the statistics to help the optimizer choose the best execution plan.
- Query Optimization: Optimizing the queries themselves can often have a greater impact on performance than simply creating indexes. Reviewing query execution plans and identifying areas for improvement, such as reducing unnecessary joins or subqueries, can help to improve overall performance.



TRIGGERS

Prevent deletion on airline table

CREATE OR REPLACE FUNCTION prevent_airline_deletion()
RETURNS trigger AS
\$\$
BEGIN
RAISE EXCEPTION 'Deletion of records in the airline table is not allowed';
RETURN NULL;
END;
\$\$
LANGUAGE plpgsql;

CREATE TRIGGER prevent_airline_deletion_trigger
BEFORE DELETE ON project.airline
FOR EACH ROW
EXECUTE FUNCTION prevent_airline_deletion();



TRIGGERS-IMPLEMENTATION

```
查询历史
321
     END:
     $$
322
     LANGUAGE plpgsql;
323
324
     CREATE TRIGGER prevent_airline_deletion_trigger
325
     BEFORE DELETE ON project.airline
     FOR EACH ROW
327
     EXECUTE FUNCTION prevent_airline_deletion();
328
329
     delete from project.airline
330
331
     where carrierno =1
332
    --stored procedure
333
     CREATE OR REPLACE PROCEDURE update_airport_name(
     airportno integer,
Data Output 消息 通知
ERROR: Deletion of records in the airline table is not allowed
CONTEXT: PL/pgSQL function prevent_airline_deletion() line 3 at RAISE
SOL 状态: P0001
```

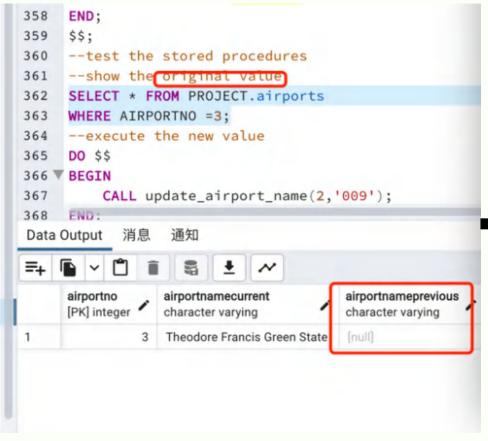


STORED PROCEDURES

```
CREATE OR REPLACE PROCEDURE update_hotel_name(
  hotelno integer,
  new_name varchar(25)
LANGUAGE plpgsql
AS $$
DECLARE
  old_name varchar(25);
BEGIN
  -- Get the current value of the "HotelNameCurrent" attribute
  SELECT HotelNameCurrent INTO old_name FROM project.HotelName WHERE HotelNameNo = hotelno;
  -- If the new value is the same as the old value, do nothing
  IF old name = new name THEN
    RETURN;
  END IF;
  -- Otherwise, update the "HotelNamePrevious" attribute with the old value
  UPDATE project.HotelName SET HotelNamePrevious = old_name WHERE HotelNameNo = hotelno;
  -- Update the "HotelNameCurrent" attribute with the new value
  UPDATE project.HotelName SET HotelNameCurrent = new_name WHERE HotelNameNo = hotelno;
END;
$$;
```



STORED PROCEDURES-RENEW



```
419 END:
420 $$;
    SELECT * FROM PROJECT.hotelname
     WHERE hotelnameno = 3
    --execute the same value
425 DO $$
         CALL update_hotel_name(2, 'Marriot');
428 END;
429 $$;
430 -- see changes
Data Output 消息 通知
手 幅 v 自 i  高 き ル
                               hotelnamecurrent
     hotelnameno , hotelnameprevious
     [PK] integer character varying
                               character varying
```

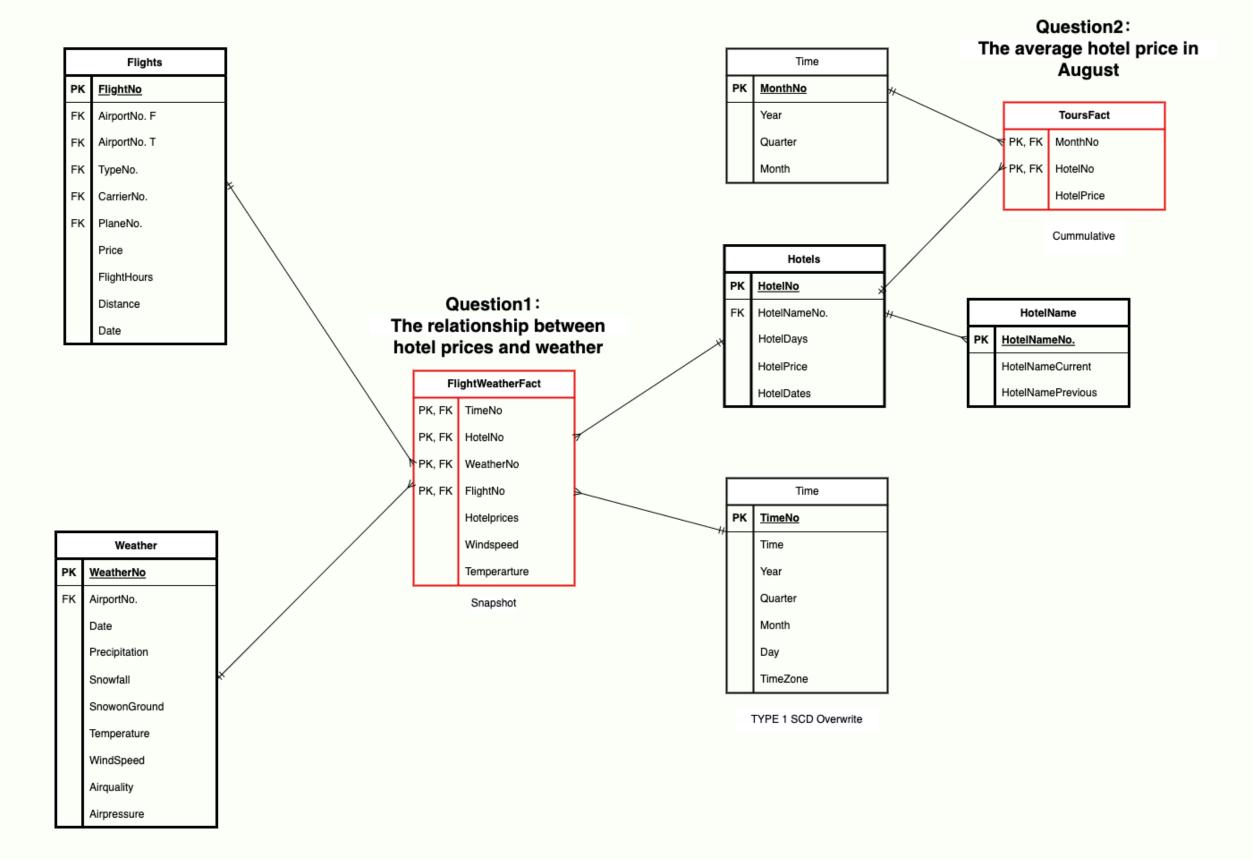
```
WHERE hotelnameno = 2
    --execute the same value
425 DO $$
426 ▼ BEGIN
          CALL update_hotel_name(2, 'Marriot');
     END;
     $$:
430 -- see changes
Data Output 消息 通知
      hotelnameno
                  hotelnameprevious
                                   hotelnamecurrent
      [PK] integer
                  character varying
                                   character varying
                  Hotel A
                                    Marriot
```

```
368
     END;
369
    --see changes
    SELECT * FROM PROJECT.airports
     WHERE airportno = 3
373
    --Dimensional&creation
374
375 CREATE TABLE project.time (
376 TIME TIMESTAMP,
Data Output 消息
=+ ⓑ ∨ ˚
                                   airportnameprevious
     [PK] integer
                 character varying
                                   character varying
                  Boston Logan
                                    Theodore Francis Green State
```

```
428
    END;
429 $$;
    --see changes
431 SELECT * FROM PROJECT.hotelname
     WHERE hotelnameno = 2
433
434
126 __Dimonsional Percetion
Data Output 消息 通知
                  hotelnameprevious
                                  hotelnamecurrent
     [PK] integer
                 character varying
                                  character varying
               2 Hotel A
```

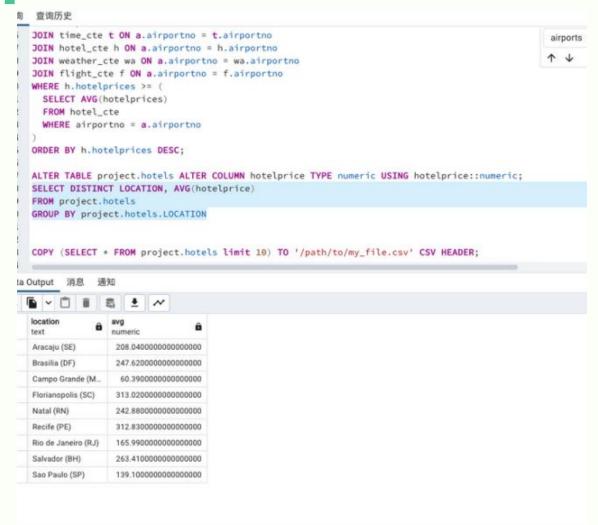


DIMENSION&FACT





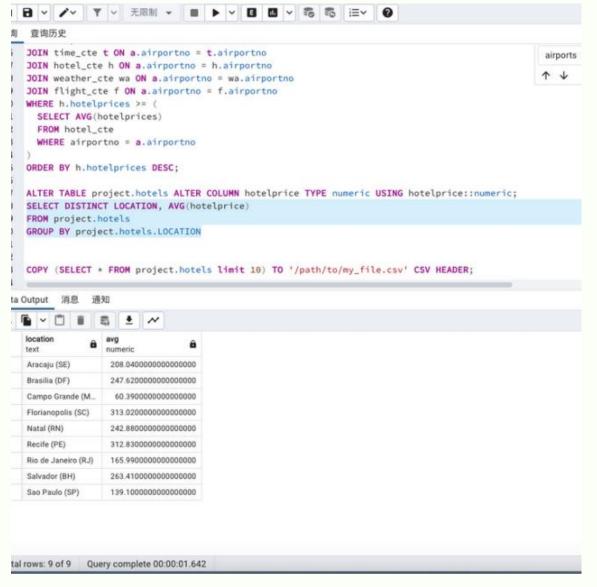
Question: What is the average hotel price in each area



COPY (SELECT DISTINCT LOCATION, AVG(hotelprice) FROM project.hotels
GROUP BY project.hotels.LOCATION) TO 'TO '/users/yuanditang/downloads/avg.csv';

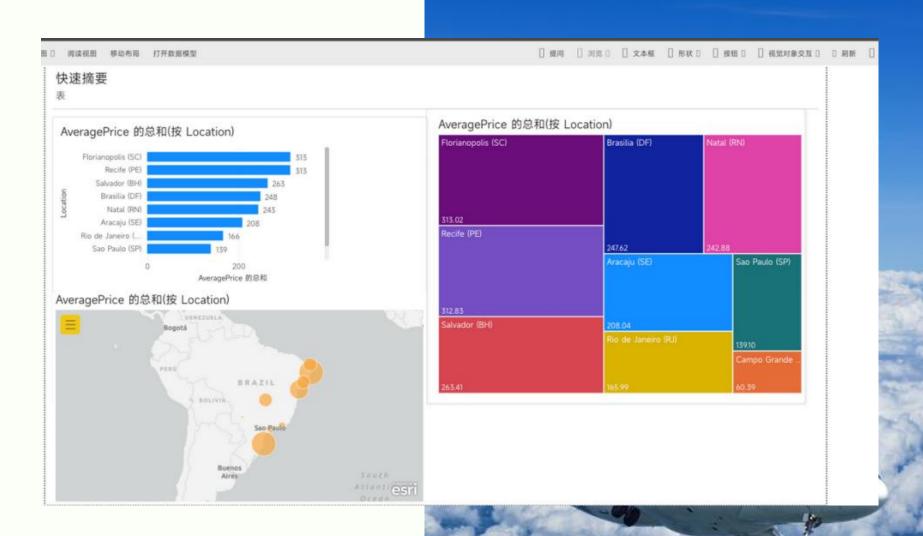


POWERBI CHARTS





COPY (SELECT DISTINCT LOCATION, AVG(hotelprice) FROM project.hotels
GROUP BY project.hotels.LOCATION) TO 'TO '/users/yuanditang/downloads/avg.csv';





Elaboration Likelihood Model (ELM): This theory proposes that there are two routes to persuasion: a central route and a peripheral route. The central route involves careful consideration and evaluation of the message, while the peripheral route relies on cues such as source credibility or emotional appeals. In airline marketing, the ELM could be used to design messages that appeal to both routes, such as highlighting the safety record of the airline (central route) while also using attractive visuals and emotional music in the marketing campaign (peripheral route).



Social Exchange Theory: This theory suggests that human relationships are based on an exchange of rewards and costs. In airline marketing, the social exchange theory could be applied by emphasizing the rewards of flying with the airline (such as comfort, convenience, and status) and minimizing the costs (such as price, inconvenience, and risk).

Uses and Gratifications Theory: This theory proposes that people use media to satisfy specific needs, such as information, entertainment, or social interaction. In airline marketing, the uses and gratifications theory could be applied by designing messages that appeal to the different needs of the target audience, such as providing information about flight schedules and amenities, showcasing the destination and cultural experiences, or promoting social connections and community through travel.



Social Learning Theory: This theory suggests that people learn new behaviors by observing and imitating others. In airline marketing, the social learning theory could be applied by featuring endorsements and testimonials from satisfied customers, celebrities, or influencers who embody the values and lifestyle associated with the airline brand. This could also be extended to social media campaigns that encourage usergenerated content and sharing of travel experiences.

