



# DATA WAREHOUSE

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# **Background**

The project is based off an aircraft company named FlyU. They hold records identified with their flights, for example, take-off times, appearance times, customers on each flight, number of trips to arrive at their destination, return flights, tail number, grievances raised, and any sort of flight delays.

The company stores utilizing Oracle just as Excel sheets. They handle this everyday information to design, deal with their client grumblings support and convey a quality assistance.

# Aims/Objectives

FlyU company aims for the following results:

- Integrate a Data Warehouse to store the information relating to FlyU.
- Deliver a quality service
- Increase customer satisfaction
- Grow the company.

### **Data Warehouse**

Before we move further into the project, it is essential to comprehend what is data warehouse. In this way, in straightforward terms, it is a kind of data management system that stores an company information(data) from at least one sources. The fundamental reason for data warehouse is to think about and investigate information for more noteworthy corporate execution. With regards to business knowledge, it is viewed as one the imperative parts as it helps uphold business choices by giving analytical techniques and a more extensive understanding into the data warehouse. It is intended to execute query and analysis on historical data derived from heterogeneous sources multiple sources.

### **PURPOSE OF DW:**

These are the main objective of implementing a Data Warehouse into a company:

### Improve quality of data:

One of the main purpose of Data Warehouse is to guarantee data quality. Any bad or error data are analysed, purified, and transformed into a useable data hence ensuring good data quality.

### Minimize inconsistent reports:

Since, inconsistent reports are mainly caused by misuse of data, and the main reason for misuse of data is disagreement or misunderstanding of the meaning or the content of data. Data Warehouse ensures that there is no disagreements or misunderstandings.

### • Integrate data from multiple sources:

Another prime objective of Data Warehouse is to make it easy for companies to integrate data from multiple sources.

### Merge historical data with current data:

As source systems do not usually keep a history of certain data, typical data warehouse objective is to store history. In data warehouse data changes in the source system are recorded, which enables historical analysis.

### OLAP:

Data Warehousing - OLAP. Online Analytical Processing Server (OLAP) is based on the multidimensional data model. It allows managers, and analysts to get an insight of the information through fast, consistent, and interactive access to information. OLAP cubes enable four basic types of multidimensional data analysis:

- Drill-down: The drill-down operation converts less-detailed data into more-detailed data
- Roll-up: Roll up is the opposite of the drill-down function as it aggregates detailed data.
- **Slice and Dice:** The slice operation creates a sub-cube by selecting a single dimension and the dice operation isolates a sub-cube by selecting several dimensions.
- Pivot: The pivot function rotates the current cube view to enable dynamic multidimensional views of data.

#### OLTP:

OLTP (Online Transactional Processing) is a category of data processing that is focused on transaction-oriented tasks. OLTP typically involves inserting, updating, and/or deleting small amounts of data in a database.

### **OLAP Vs OLTP:**

The differences between OLAP and OLTP:

OLAP	OLTP
The primary objective is data analysis.	The primary objective is data processing.
OLAP can be used for all type of business analysis needs which includes planning, budgeting, forecasting, and analysis	OLTP is useful to administer day to day transactions of an organization.
OLAP uses data warehouse technique where it can integrate different data sources for building a secure database.	OLTP uses traditional DBMS

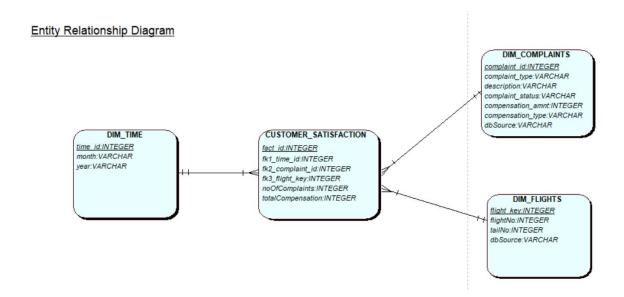
### **KPI 2: Ensure customer satisfaction**

### Task 1

### Reports:

- Identify the no of customer complaints per complaint type per flight registered in the year 2017.
- The number of complains per month.
- Total amount of compensation given per month.
- Total amount of compensation given per month per flight.
- Total number of complaints per month per complaint type.

### **STAR SCHEMA:**



Star schema is one of the crucial components of Data Warehouse. In the star schema, there is a fact table "Customer Satisfaction" surrounded by many dimensions "Dim Flights", "Dim Complaints" and "Dim Time". The fact table has the foreign keys from the different dimensions as well as all the measures that are relevant to the project reports.

### DATA DICTIONARY:

Data dictionary provides a clear picture about the contents, format, and structure of a database and the relationship between its elements as well as any bad data that need to be transformed.

# DIM\_TIME

Star Schema Table	Attribute Name	Data Type	Key	DQ Source	Data Mapping	Data quality Issues	Transformation
				Automatically	11 0		Create a sequence
	timeID	Integer	Yes	generated as	n/a	n/a	timeid_seq to generate
			163	a primary key			primary keys
	year	Number	No	FlyU_flights	n/a	n/a	n/a
	month	Number	No	FlyU_flights	n/a	n/a	n/a
	day	Number	No	FlyU_flights	n/a	n/a	n/a
							Create a sequence
DIM_TIME				Should be			quarter_seq to generate
	quarter	VARCHAR	No	generated	n/a	n/a	quarterly dates
	The dim_t	time table ho	olds th	e intervals of tim	e for which	the data will	be held. It is held at year,
Definition:	month, da	ay and quart	erly le	vel meaning.			
Notes:							

# DIM\_FLIGHTS

Star Schema Table	Attribute Name	Data Type	Key	DQ Source	Data Map ping	Data quality Issues	Transformation			
				Automatically			Create a sequence			
	flightKey	INTEGER	Yes	generated as	n/a	n/a	flight_seq to generate			
				surrogate key			primary keyS			
	flightNo	INTEGER	No	FlyU_flights	n/a	n/a	n/a			
	TailNo	INTEGER	No	FlyU_flights	n/a	n/a	n/a			
		VARCHAR					Create a sequence			
DIM_FLIGHTS				Should be			SOURCE_seq to generate			
	dbSource		No	generated	n/a	n/a	Quarterly dates			
Definition:	The dim_F	The dim_FLIGHTS table holds the data related to all the flights.								
Notes:										

# DIM\_COMPLAINTS

Star Schema Table	Attribute Name	Data Type	Key	DQ Source	Data Mapping	Data quality Issues	Transformation
	Complaint Key	INTEGER	Yes	Automaticall y generated as surrogate key	n/a	n/a	Create a sequence complain_seq to generate primary keys
	Complaint Id	INTEGER	No	FlyU_flights	n/a	n/a	n/a
		VARCHAR			Null value	Some complaints are missing the compensation type	Will need to add the missing compensation type
	Complaint Type		No	FlyU_flights	Inconsist ent value	Complaint types have irregural values. Eg: A,B,C for cancellation	Will need to transform all the irregular values to 'C' for cancellation and 'L' for late.
	Descriptio n	VARCHAR	No	FlyU_flights	Null Value	Some complaints are missing the description	Will need to add the missing description
	Complaint status	VARCHAR	No	FlyU_flights	n/a	n/a	n/a
DIM_	compensa tion_amnt	INTEGER	No	FlyU_flights	n/a	n/a	n/a
COMPLAINTS	compensa tion_type	VARCHAR	No	FlyU_flights	Null value	Some complaints are missing the compensation type	Will need to mention the compensation type
	dbSource	VARCHAR	No	Should be generated	n/a	n/a	Create a sequence SOURCE_seq to generate quarterly dates
Definition:	The dim_co	mplaint tabl	e hold	s all the data re	lated to the	e customer complain	ts.
Notes:							

■ FACT\_CUSTOMER\_SATISFACTION

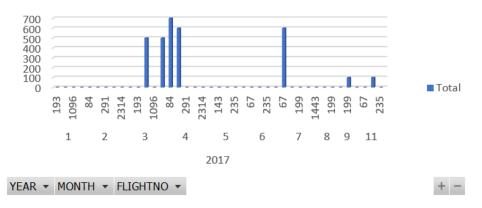
Star Schema Table	Attribute Name	Data Type	Key	DQ Source	Data Mapping	Data quality Issues	Transformation		
	FactId	INTEGER	Yes	Automatical ly generated as primary key	n/a	n/a	Create a sequence fact_seq to generate primary keys		
	Complaint Key	INTEGER	No	FlyU_flights	n/a	n/a	n/a		
- 54 OT OU	Compensation Key	INTEGER	No	FlyU_flights	n/a	n/a	n/a		
FACT_CU STOMER	TimeId	INTEGER	No	FlyU_flights	n/a	n/a	n/a		
_SATISFA	flightKey	INTEGER	No	FlyU_flights	n/a	n/a	n/a		
CTION	noOfComplaint	INTEGER	No	Should be generated	n/a	n/a	Create a sequence SOURCE_seq to generate quarterly dates		
	total Compensation	INTEGER	No	Should be generated	n/a	n/a	Create a sequence SOURCE_seq to generate quarterly dates		
Definition:	The dim_complaint table holdsall the data related to the customer complaints.								
Notes:									

### ANALYTICAL REPORT:

■ EXCEL REPORT

### Sum of TOTALCOMPENSATION

# Total Compensation Per Month Per Flight



### DRILL DOWN

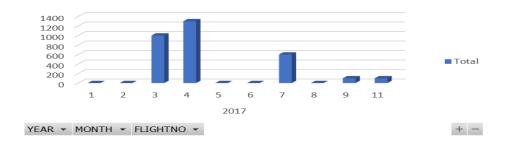


2017

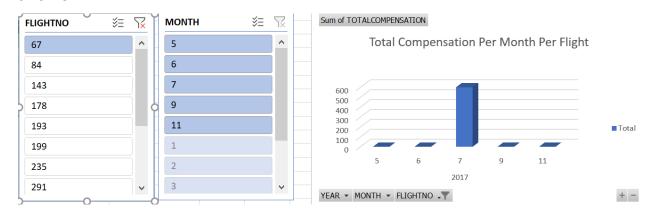
### DRILL UP



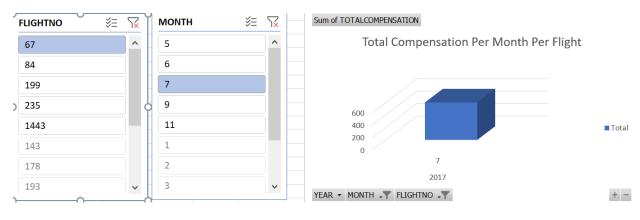
Total Compensation Per Month Per Flight



### SLICING



### DICING

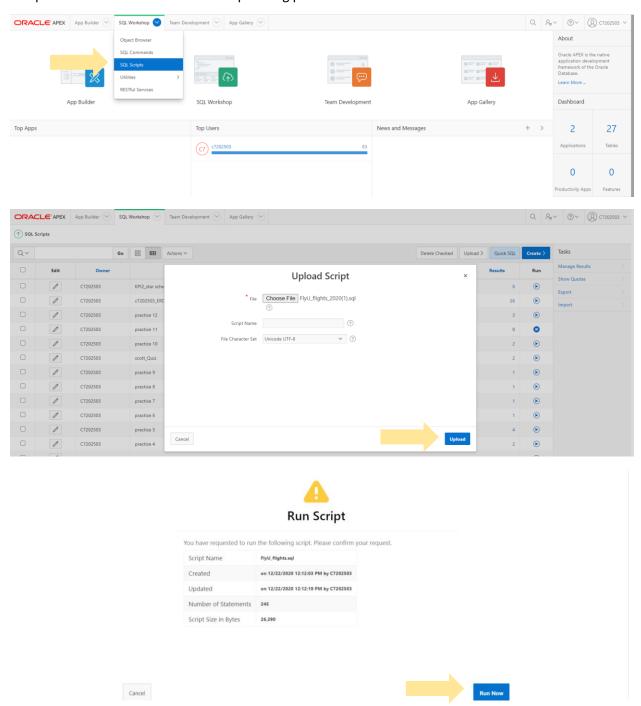


# Task 2

### **STAGING AREA SETUP:**

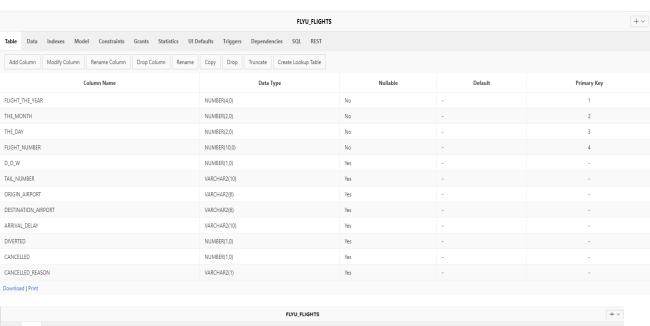
In the staging area setup, we have extracted data from the data source: flyU\_flights as we only need data related to the customer satisfaction.

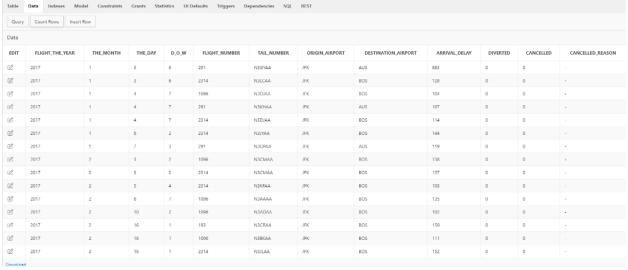
The pictures below demonstrate the uploading process of the data source.



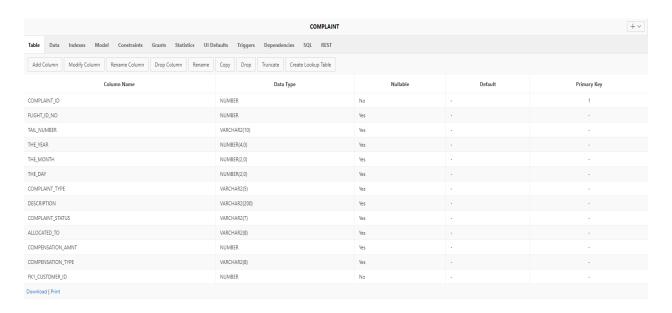
Number ↑L	Elapsed	Statement	Feedback	Rows
1	0.02	DROP TABLE FlyU_flights CASCADE CONSTRAINTS	Table dropped.	0
2	0.01	DROP TABLE customer CASCADE CONSTRAINTS	Table dropped.	0
3	0.02	DROP TABLE complaint CASCADE CONSTRAINTS	Table dropped.	0
4	0.01	CREATE TABLE customer( customer_id INTEGER NOT NULL, custo	Table created.	0
S	0.01	CREATE TABLE complaint( complaint_id INTEGER NOT NULL PRIMA	Table created.	0
6	0.01	CREATE TABLE FlyU_flights ( flight_the_year NUMBER(4), t	Table created.	0
7	0.01	ALTER TABLE FlyU_flights ADD CONSTRAINT pk1_flights PRIMARY	Table altered.	0
8	0.01	ALTER TABLE complaint ADD CONSTRAINT fk1_complaint_to_custom	Table altered.	0
9	0.01	ALTER TABLE complaint ADD CONSTRAINT fk1_complaint_to_flight	Table altered.	0
10	0.01	INSERT INTO Customer VALUES (10, 'NY10', 'BUSINESS', 'IBM',	1 row(s) inserted.	1
11	0.00	INSERT INTO Customer VALUES (101, "NY101", "BUSINESS", "Goog	1 row(s) inserted.	3
12	0.00	INSERT INTO Customer VALUES (102, "NY102", "BUSINESS", "Amaz	1 row(s) inserted.	1
13	0.00	INSERT INTO Customer VALUES (103, 'NY103', 'BUSINESS', 'Face	1 row(s) inserted.	1
14	0.01	INSERT INTO Customer VALUES (104, 'NY104', 'BUSINESS', 'HP',	1 row(s) inserted.	1
15	0.00	INSERT INTO Customer VALUES (105, 'NY105', 'BUSINESS', 'SkyB	1 row(s) inserted.	1
Download				
	245 Statements Processed	245 Successful	O With Errors	row(s) 1 - 15 of 245 Next ▶
Q c7202503 @ c7202503 @	) en	Copyright © 1999, 2019, Oracle, All rights re	served.	Application Express 19.1.0.00.15

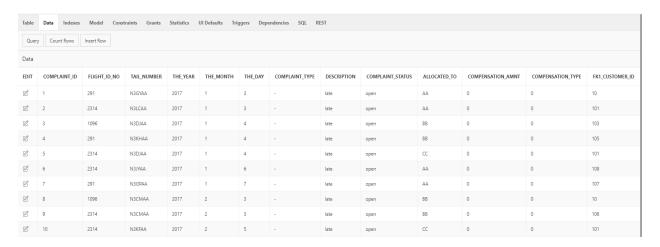
# FlyU\_flights



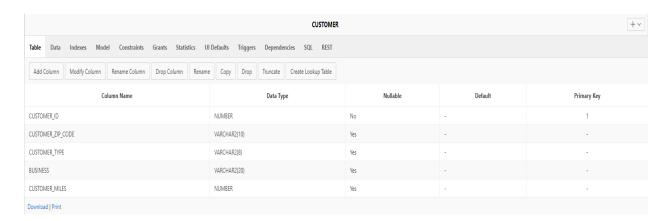


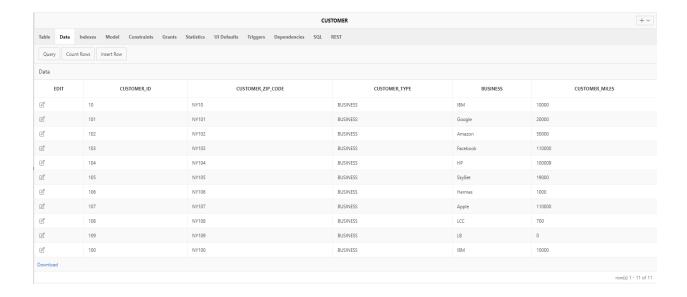
### Complaint





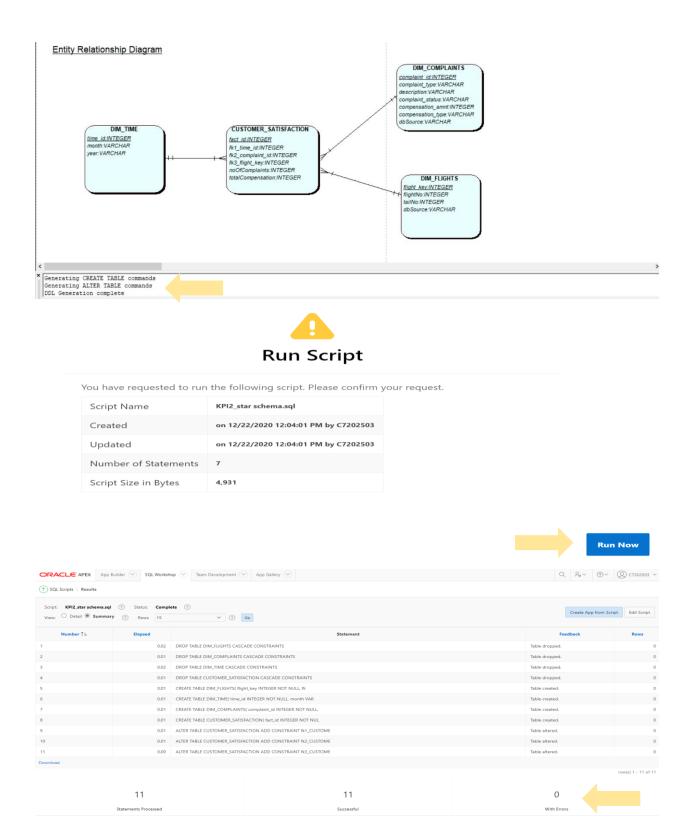
### Customer





### **STAR SCHEMA SETUP:**

The star schema that has been generated is converted into a sql script and uploaded into database using the following steps:



### Task 3

# ETL(Extraction, Transformation, Load)

ETL is short for extract, transform, load. It is an important data warehouse(DW) tool to pull data from multiple database and place it into another database. ETL is one of the key tool for today's business intelligence (BI) processes and systems. It is the process which facilitates analytical tools and important business insights.

#### **EXTRACTION:**

Extraction is the process of migrating data from multiple and different types of sources. In this case, we have already extracted the data, now we are going to transfer the data into a staging area.

Firstly, the stage tables are created along with their primary keys, foreign keys, attributes, etc. There are two staging tables created based on our star schema design and they are:

### **Stage Complaints:**

### Stage Flights:

### **Data Insertion:**

In the figure below, the staging is done using STG\_DATA\_INSERT procedure which uses loop to insert data into all the staging tables. This ensures that the data are accurate as it is sent based on the foreign key.

```
INSERT DATA INTO STAGING TABLES
______
create or replace procedure STG_DATA_INSERT
next NUMBER;
begin
       FOR i IN (SELECT * FROM FlyU Flights) LOOP
                next := STG FLIGHT SEO.nextval;
                INSERT INTO STG_FLIGHTS(flight key, flight no, year, month, day, tailNo, dbSource)
                VALUES(next, i.flight number, i.flight the year, i.the month, i.the day, i.tail number, 'flyU_flights_2020');
       FOR j IN (SELECT * FROM complaint WHERE flight id no=i_flight number and the year=i_flight the year and the month and the day=i_the day) LOOP
               INSERT INTO STG_COMPLAINTS(complaint id, complaint type, description, complaint status, compensation_ammt, compensation_type, flight Key, dbSource)
               VALUES(<u>j.complaint id</u>, <u>j.complaint type</u>, <u>j.description</u>, <u>j.complaint status</u>, <u>j.compensation amnt</u>, <u>j.compensation type</u>, next,
               'flyU_flights_2020');
       END LOOP;
   END LOOP;
END;
begin
STG_DATA_INSERT;
```

### After that, the tables look like following:

	STG_COMPLAINTS + V										
Table	Table Data Indexes Model Constraints Grants Statistics UI Defaults Triggers Dependencies SQL REST										
Query	Count Rows Inse	rt Row									
Data											
EDIT	COMPLAINT_ID	COMPLAINT_TYPE	DESCRIPTION	COMPLAINT_STATUS	COMPENSATION_AMNT	COMPENSATION_TYPE	FLIGHT_KEY	DBSOURCE			
Ø	1	-	late	open	0	0	1	flyU_flights_2020			
C	2	-	late	open	0	0	2	flyU_flights_2020			
C	3	-	late	open	0	0	3	flyU_flights_2020			
Ø	4	-	late	open	0	0	4	flyU_flights_2020			
C	5	-	late	open	0	0	5	flyU_flights_2020			
C	6	-	late	open	0	0	6	flyU_flights_2020			
C	7	-	late	open	0	0	7	flyU_flights_2020			

	STG_FLIGHTS										
Table Data Indexes Model Constraints Grants Statistics UI Defaults Triggers Dependencies SQL REST											
Query Cour	Query Count Rows Insert Row										
Data	Data										
EDIT	FLIGHT_KEY	FLIGHT_NO	YEAR	MONTH	DAY	TAILNO	DBSOURCE				
	1	291	2017	1	3	N3GYAA	flyU_flights_2020				
<b>C</b>	2	2314	2017	1	3	N3LCAA	flyU_flights_2020				
<b>C</b>	3	1096	2017	1	4	N3DJAA	flyU_flights_2020				
C	4	291	2017	1	4	N3KHAA	flyU_flights_2020				
C	5	2314	2017	1	4	N3DJAA	flyU_flights_2020				
C	6	2314	2017	1	6	N3JYAA	flyU_flights_2020				
ď	7	291	2017	1	7	N3GPAA	flyU_flights_2020				

### TRANSFORMATION:

Transform is the process of converting the extracted data from its previous form into the form it needs to be in so that it can be placed into another database.

In this part of the ETL process, we identify all the bad and good data determined in the data dictionary, create good and bad tables and populate it, reclean bad data and transfer the bad data into the good tables.

### **COMPLAINT TABLE TRANSFORMATION**

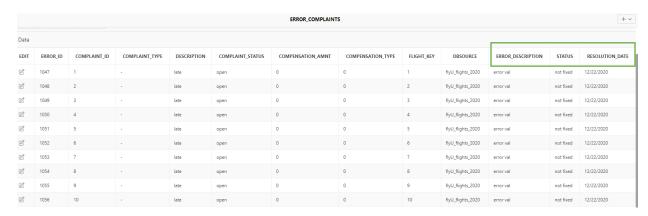
Firstly, two tables where clean and error data will be segregated are created.

```
CREATING ERROR DATA TABLE FOR COMPLAINTS TABLE
CREATING CLEAN DATA TABLE FOR COMPLAINTS TABLE
-----
                                                         truncate table error complaints;
truncate table clean complaints;
                                                        DROP table error complaints CASCADE CONSTRAINTS;
DROP table clean complaints CASCADE CONSTRAINTS;
                                                        create table error complaints
create table clean complaints
                                                                "ERROR ID" NUMBER(*,0) NOT NULL PRIMARY KEY,
                                                                "COMPLAINT_ID" NUMBER(*,0) NOT NULL ENABLE,
        "COMPLAINT_ID" NUMBER(*,0) NOT NULL PRIMARY KEY,
                                                                "COMPLAINT_TYPE" VARCHAR2(5),
                                                                "DESCRIPTION" VARCHAR2(20),
        "COMPLAINT TYPE" VARCHAR2(5),
                                                                "COMPLAINT STATUS" VARCHAR2(7),
        "DESCRIPTION" VARCHAR2(20),
                                                                "COMPENSATION_AMNT" NUMBER(*,0),
       "COMPLAINT STATUS" VARCHAR2(7),
                                                                "COMPENSATION_TYPE" VARCHAR2(8),
        "COMPENSATION AMNT" NUMBER(*,0),
                                                               "FLIGHT KEY" INTEGER NOT NULL,
                                                               "DBSOURCE" VARCHAR2(17),
        "COMPENSATION_TYPE" VARCHAR2(8),
                                                               "ERROR_DESCRIPTION" VARCHAR2(15),
        "FLIGHT_KEY" INTEGER NOT NULL,
                                                                "STATUS" VARCHAR2(10),
        "DBSOURCE" VARCHAR2(17)
                                                                "RESOLUTION DATE" DATE
);
                                                         );
```

Now, that the tables are created, we are going to identify the clear and error data using procedure and populate the tables.

```
create or replace procedure check_complaints_data(no_rows OUT NUMBER)
BEGIN
no_rows := 0;
INSERT INTO error_complaints
SELECT ERROR_COMPLAINTS_SEQ. nextval, c.COMPLAINT_ID, c.COMPLAINT_TYPE, c.DESCRIPTION, c.COMPLAINT_STATUS, c.COMPENSATION_ANNT, c.COMPENSATION_TYPE, c.FLIGHT_KEY, c.DBSOURCE, 'error val', 'not fixed', sysdate
FROM STG_COMPLAINTS c
WHERE COMPLAINT_TYPE IS NULL
OR DESCRIPTION IS NULL
OR COMPLAINT_STATUS IS NULL
OR COMPENSATION_AMNT IS NULL
OR COMPENSATION_TYPE IS NULL
OR description ='cancelled' AND COMPLAINT_TYPE <>'C'
no_rows := TO_CHAR(SQL%RowCount);
DELETE FROM clean_complaints;
INSERT INTO clean_complaints
SELECT * FROM STG_COMPLAINTS
WHERE COMPLAINT_TYPE IS NOT NULL
AND DESCRIPTION IS NOT NULL
AND COMPLAINT STATUS IS NOT NULL
AND COMPENSATION AMNT IS NOT NULL
AND COMPENSATION_TYPE IS NOT NULL
AND description ='cancelled' AND COMPLAINT TYPE ='C'
```

Then, the bad table looks as shown in the picture. Here, all the bad data will have error description, status, and resolution date.



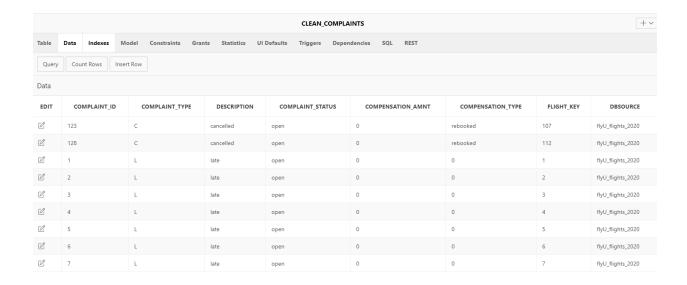
As we have identified the bad data, we will now reclean the error data and make it useable.

This is done by using a procedure as shown below.

After recleaning, since the bad and good data are segregated, we will now bring them in the same table using a merge statement.

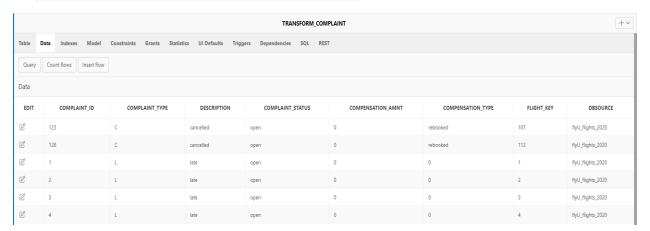
```
create or replace procedure RECLEANED_COMPLAINT_DATA_MERGE
begin
MERGE INTO clean complaints C
USING error complaints B
 ON (C.COMPLAINT_ID = B.COMPLAINT_ID)
WHEN MATCHED THEN
C.COMPLAINT_TYPE = B.COMPLAINT_TYPE,
C.DESCRIPTION=B.DESCRIPTION,
C.COMPLAINT_STATUS=B.COMPLAINT_STATUS,
C.COMPENSATION_AMNT = B.COMPENSATION_AMNT,
C.COMPENSATION_TYPE = B.COMPENSATION_TYPE,
C.FLIGHT_KEY = B.FLIGHT_KEY,
C.DBSOURCE=B.DBSOURCE
WHEN NOT MATCHED THEN
 INSERT VALUES
(B.COMPLAINT_ID, B.COMPLAINT_TYPE, B.DESCRIPTION, B.COMPLAINT_STATUS, B.COMPENSATION_AMNT, B.COMPENSATION_TYPE, B.FLIGHT_KEY, B.DBSOURCE);
END;
RECLEANED_COMPLAINT_DATA_MERGE;
end;
```

The figure below shows the reclean data as well as the clean data from before.



Finally, we will move the data from the cleaning stage to the transformation table so that it can be loaded into the dimensional model.

```
--TRANSFORM_COMPLAINT-
DROP TABLE TRANSFORM COMPLAINT CASCADE CONSTRAINTS;
TRUNCATE TABLE TRANSFORM_COMPLAINT;
CREATE TABLE TRANSFORM_COMPLAINT
COMPLAINT_ID NUMBER NOT NULL PRIMARY KEY,
"COMPLAINT_TYPE" VARCHAR2(5),
"DESCRIPTION" VARCHAR2(200),
"COMPLAINT_STATUS" VARCHAR2(7),
"COMPENSATION_AMMIT" NUMBER,
"COMPENSATION_TYPE" VARCHAR2(20),
"FLIGHT_KEY" INTEGER NOT NULL,
"DBSOURCE" VARCHAR2(17)
-- PROCEDURE TO TRANSFORM GOOD COMPLAINT TABLE
create or replace procedure TRANSFORM_COMPLAINT_TABLE
begin
DELETE FROM TRANSFORM_COMPLAINT;
INSERT INTO TRANSFORM_COMPLAINT (SELECT * FROM CLEAN_COMPLAINTS);
END;
TRANSFORM_COMPLAINT_TABLE;
end:
```



### **FLIGHTS TABLE TRANSFORMATION**

Firstly, two tables where clean and error data will be segregated are created.

```
CREATING BAD DATA TABLE FOR FLIGHTS TABLE

DROP table error flights (
"BAD_ID" NUMBER(*,0) NOT NULL PRIMARY KEY,
   "FLIGHT_KEY" NUMBER(*,0),
   "FLIGHTNO" NUMBER(10,0),
   "YEAR" NUMBER(2,0),
   "DAY" NUMBER(2,0),
   "TAILNO" VARCHAR2(10),
   "BSOURCE" VARCHAR2(17),
   "ERROR_DESCRIPTION" VARCHAR2(15),
   "STATUS" VARCHAR2(10),
   "RESOLUTION_DATE" DATE
);

CREATING GOOD DATA TABLE FOR FLIGHTS TABLE

DROP table clean flights CASCADE CONSTRAINTS;
   Create table clean flights
(
FLIGHT_KEY NUMBER NOT NULL PRIMARY KEY,
   "FLIGHTNO" NUMBER(10,0),
   "YEAR" NUMBER(4,0),
   "MONTH" NUMBER(2,0),
   "DAY" NUMBER(2,0),
   "TAILNO" VARCHAR2(10),
   "DAY" NUMBER(2,0),
   "TAILNO" VARCHAR2(10),
   "DBSOURCE" VARCHAR2(17)
);
```

Now, that the tables are created, we are going to identify the clear and error data using procedure, and populate the tables.

```
create or replace procedure check_flights_data(no_rows OUT NUMBER)
BEGIN
no_rows := 0;
INSERT INTO error_flights
SELECT ERROR_FLIGHTS_SEQ.nextval, F.FLIGHT_KEY, F.FLIGHT_NO, F.YEAR, F.MONTH, F.DAY, F.TAILNO, F.DBSOURCE, 'null val', 'not fixed', sysdate
FROM STG_FLIGHTS F
WHERE FLIGHT NO IS NULL
OR YEAR IS NULL
OR MONTH IS NULL
OR DAY IS NULL
OR TAILNO IS NULL);
no_rows := TO_CHAR(SQL%RowCount);
INSERT INTO error_flights
SELECT ERROR_FLIGHTS_SEQ.nextval, F.FLIGHT_KEY, F.FLIGHT_NO, F.YEAR, F.MONTH, F.DAY, F.TAILNO, F.DBSOURCE, 'inconsistent', 'not fixed', sysdate
FROM STG FLIGHTS F
WHERE F.YEAR='17');
no_rows := no_rows + TO_CHAR(SQL%RowCount);
DELETE FROM clean_flights;
INSERT INTO clean_flights
SELECT * FROM STG_FLIGHTS
WHERE FLIGHT_NO IS NOT NULL
OR YEAR IS NOT NULL
OR MONTH IS NOT NULL
OR DAY IS NOT NULL
OR TAILNO IS NOT NULL
AND YEAR <>'17');
END:
```

Then, the bad table looks as shown in the picture. Here, all the bad data will have error description, status, and resolution date.

	ERROR_FLIGHTS +										+~
Table D	Table Data Indexes Model Constraints Grants Statistics UI Defaults Triggers Dependencies SQL REST										
Query	Query Count Rows Insert Row										
Data											
EDIT	BAD_ID	FLIGHT_KEY	FLIGHTNO	YEAR	MONTH	DAY	TAILNO	DBSOURCE	ERROR_DESCRIPTION	STATUS	RESOLUTION_DATE
Ø.	3	8	1096	17	2	3	N3CMAA	flyU_flights_2020	inconsistent	not fixed	12/22/2020
Ø	4	9	2314	17	2	3	N3CMAA	flyU_flights_2020	inconsistent	not fixed	12/22/2020
Download											

row(s) 1 - 2 of 2

As we have identified the bad data, we will now reclean the error data and make it useable.

This is done by using a procedure as shown below.

```
Reclean bad data in FLIGHT bad table--

create or replace procedure reclean flight bad data(py rows OUT NUMBER)

IS

BEGIN

PY_FOWS := 0;

update error flights set YEAR='2017', STATUS ='fixed', RESOLUTION_DATE = sysdate where YEAR='17';

py_rows := py_rows + TO_CHAR(SQL%RowCount);

end;

EXECUTE Compensation Table PROCEDURE

DECLARE

NOMBER(5,2);

begin

reclean flight bad data(noOfRows);

DBMS_OUTPUT.PUT_LINE('No of bad data cleaned: '||noOfRows);

END;

Results

Explain Describe Saved SQL History

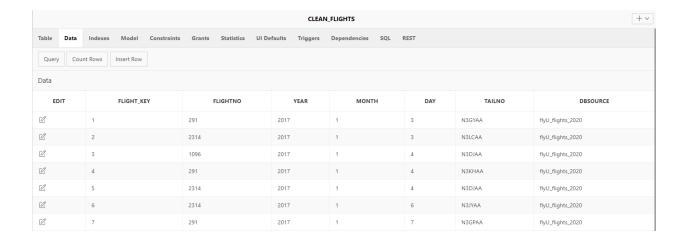
No of bad data cleaned: 2

Statement processed.
```

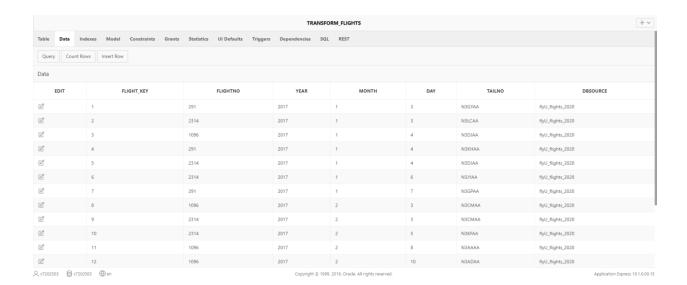
Since the bad and good data are segregated, we will now bring them in the same table using a merge statement.

```
create or replace procedure RECLEANED_FLIGHT_DATA_MERGE
as
begin
MERGE INTO clean_flights C
USING error flights B
ON (c_flight key = B_flight key)
WHEN MATCHED THEN
UPDATE SET
C.FLIGHTNO = B.FLIGHTNO,
C.YEAR = B.YEAR,
C.MONTH = B.MONTH,
C.DAY = B.DAY,
C.TAILNO=B.TAILNO,
C.DESOURCE=B.DBSOURCE
WHEN NOT MATCHED THEN
INSERT VALUES
(B_flight key, B.FLIGHTNO, B.YEAR, B.MONTH, B.DAY, B.TAILNO, B.DBSOURCE);
END;
begin
RECLEANED_FLIGHT_DATA_MERGE;
end;
```

The figure below shows the reclean data as well as the clean data from before.



# Finally, we will move the data from the cleaning stage to the transformation table so that it can be loaded into the dimensional model.



### LOAD:

Load is the process of writing the data into the target database. In this part of the ETL process, we load all the data from the transformation table into the star schema design we upload above in task 1.

The insertion is done using procedure which utilizes loop to insert data with unique constraints into the dimension tables.

```
CREATE OR REPLACE PROCEDURE DIM_TABLE_INSERT

as

v_count NUMBER;

BEGIN

FOR i IN (SELECT * FROM transform flights) LOOP

INSERT INTO dim flights(flight key, flightno, tailno, dbsource) VALUES(i.flight key, i.flightno, i.dbsource);

SELECT count(*) INTO v_count FROM dim time WHERE year=i,year and month=i,month;

IF v_count=0 then

INSERT INTO dim time(time_id, year, month) VALUES(DIM_TIME_SEQ.nextval, i,year, i,month);
end if;
FOR j IN (SELECT * FROM transform complaint WHERE flight key=i,flight key) LOOP

INSERT INTO dim complaints(COMPLAINT_ID, cOMPLAINT_TYPE, DESCRIPTION, COMPLAINT_STATUS, COMPENSATION_AMNT, COMPENSATION_TYPE, DBSOURCE)

VALUES( j.COMPLAINT_ID, j.COMPLAINT_TYPE, j.DESCRIPTION, j.COMPLAINT_STATUS, j.COMPENSATION AMNT, j.COMPENSATION TYPE, j.DRSOURCE );
END LOOP;

END LOOP;

END;

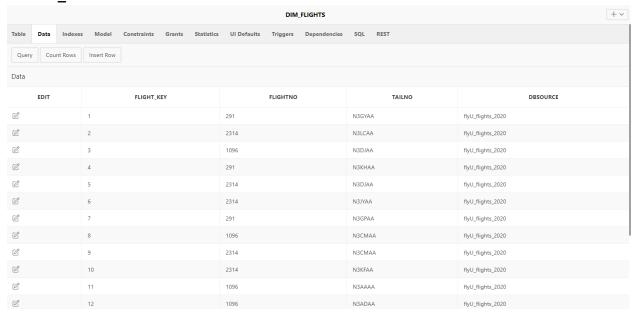
INSERT INTO dim time(time_id, year, month) VALUES(DIM_TIME_SEQ.nextval, 2017, 10);
BEGIN |

DIM_TABLE_INSERT;
END;
```

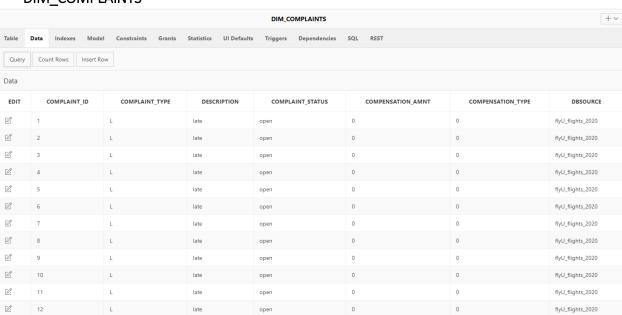
### DIM\_TIME

			DIN	I_TIME	+
Table Data	Indexes Model	Constraints Grants	Statistics UI Defaults Triggers	Dependencies SQL REST	
Query	unt Rows Insert Row				
Data					
	EDIT		TIME_ID	YEAR	MONTH
ď		14		2017	10
		15		2017	1
C		16		2017	2
C		17		2017	3
C		18		2017	4
C		19		2017	5
ď		20		2017	6
<b>C</b>		21		2017	7
<b>C</b>		22		2017	8
C		23		2017	9
C		24		2017	11
<b>C</b>		25		2017	12

# DIM\_FLIGHTS



# DIM\_COMPLAINTS



### CUSTOMER SATISFACTION (FACT TABLE)

The data insertion in fact table is done using procedure which utilizes the merge statement to insert data with unique constraints into the dimension tables.

```
CREATE OR REPLACE PROCEDURE FACT_TABLE_INSERT
MERGE INTO CUSTOMER_SATISFACTION CS
USING (SELECT t.time_id, c.complaint_ID, f.flight_key, COUNT(c.complaint_id) AS NOOFCOMPLAINTS, SUM(c.compensation_amot) as TOTALCOMPENSATION
FROM <u>transform</u> flights f
JOIN transform complaint c ON(f.flight key=c.flight key)
JOIN <u>dim time</u> t ON(<u>f.year=t.year</u> AND <u>f.month</u> = <u>t.month</u>)
GROUP BY <u>t.time_id</u>, <u>f.flight_key</u>, <u>c.complaint_id</u>) d
ON(cs.FK1_TIME_ID=<u>d.time_id</u> AND cs.FK2_FLIGHT_KEY=<u>d.flight_key</u> AND cs.FK3_COMPLAINT_ID=<u>d.complaint_id</u>)
WHEN MATCHED THEN
        UPDATE SET
        cs.NOOFCOMPLAINTS=d.NOOFCOMPLAINTS,
        cs.TOTALCOMPENSATION = d.TOTALCOMPENSATION
WHEN NOT MATCHED THEN
    INSERT
    VALUES (FACT_SEQ.NEXTVAL, d_time_id, d_flight_key, d_complaint_id, d_NOOFCOMPLAINTS, d_TOTALCOMPENSATION);
FACT_TABLE_INSERT;
```

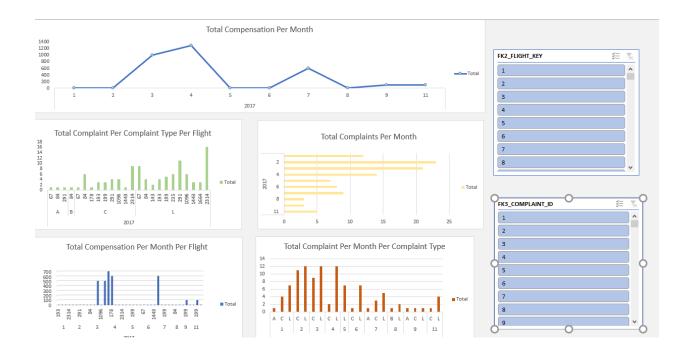
	CUSTOMER_SATISFACTION								
Table D	Table Data Indexes Model Constraints Grants Statistics UI Defaults Triggers Dependencies SQL REST								
Query	Query Count Rows Insert Row								
Data									
EDIT	FACT_ID	FK1_TIME_ID	FK2_FLIGHT_KEY	FK3_COMPLAINT_ID	NOOFCOMPLAINTS	TOTALCOMPENSATION			
ď	1	15	1	1	1	0			
ď	2	15	2	2	1	0			
ď	3	15	3	3	1	0			
Ø	4	15	4	4	1	0			
ď	5	15	5	5	1	0			
<b>U</b>	6	15	6	6	1	0			
Ø	7	15	7	7	1	0			
ď	8	16	8	8	1	0			
Ø	9	16	9	9	1	0			
ď	10	16	10	10	1	0			
<b>C</b>	11	16	11	11	1	0			
ď	12	16	12	12	1	0			

# Task 4

### Data Analysis:

### Dashboard:

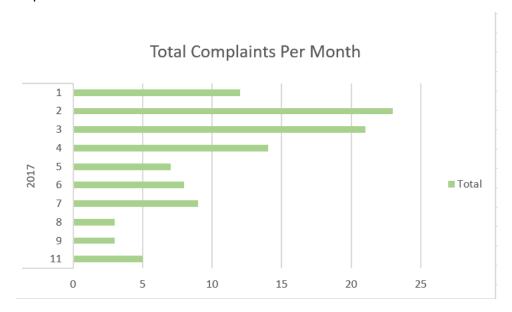
This dashboard facilitates the company to make multi-dimensional analysis based on flights, complaints and time.



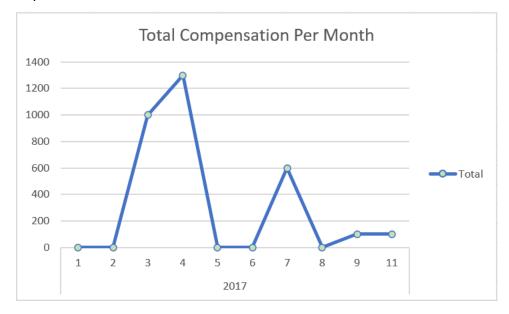
### Report 1



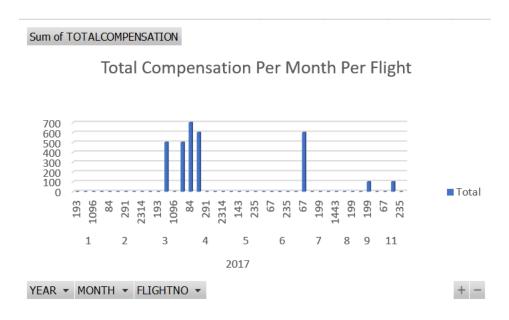
# Report 2



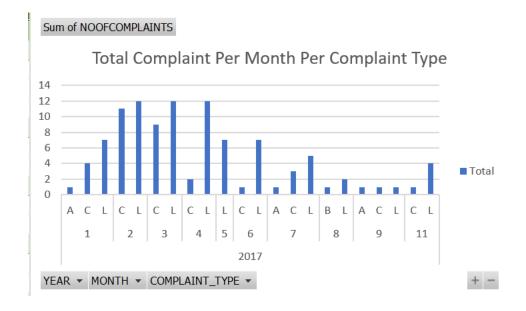
# Report 3



# Report 4



### Report 5



### Task 5

### **Data Warehouse Approaches**

In Data Warehouse, there are two major approaches when it comes to designing. They are Inmon Method and Kimball Method

With regards to effective corporate execution, it is urgent to decide the fitting methodology as per the necessity of the project. These assists cut with bringing down project cost just as save a ton of time. The two methods have their own points of interest and separating factors, so figuring out which strategy to utilize decides the fate of the company.

### **Bill Inmon's Method**

The Bill Inmon's architecture, otherwise called the top-down design, arranges information utilizing ER modelling. In this architecture, a normalized data model is planned before the dimensional data marts where all the necessary data are made from the data warehouse. The strategy authorizes information distribution centre as a concentrated data warehouse where it catches the detailed information at the most reduced degree of detail henceforth, acquiring the name, detailed data warehouse. Thus, it gives a legitimate structure to conveying business insight all things considered at the focal point of the corporate information factory (CIF). Just, it is beginning with building a major, brought together endeavour data warehouse where all accessible information from transaction system is united into a subject-situated, incorporated, time-variation that helps decision making.

The Inmon design approach uses the normalized form for building entity structure, avoiding data redundancy as much as possible. This results in clear identification of business requirements and improving any data irregularities.

### Advantages of Bill Immon's Method:

The Inmon architecture offers the following advantages:

- The data warehouse acts as a centralized unit for the entire company, where data from multiple sources can be integrated.
- This approach data warehouse process is less likely to result failure as it avoids data redundancy as much as possible resulting in relatively less data irregularities.
- As the top-down model represents data at a very lowest level of detail, making decision making and analytical process simpler.
- This approach is greatly flexible, as it is easier to update the data warehouse in case there is any change in the data sources, time, business requirements, etc.
- It can handle diverse enterprise-wide reporting requirements.

### Disadvantages of Bill Immon's Method:

The Inmon architecture offers the following disadvantages:

- It can be susceptible to more complexity because over time, multiple tables are added to the data warehouse.
- It can be expensive in terms of hiring resources skilled in data science.
- The initial setup and delivery can take a lot of time.
- Additional ETL operation is required since data marts are created after the creation of the data warehouse.
- This approach requires experts to effectively manage a data warehouse.

### **Ralph Kimball Method**

The Ralph Kimball architecture, otherwise called as bottom-up design of Data Warehouse, forms data marts first based on the business prerequisites. In this design, the key business questions and the key business measures are recognized before the essential information sources are assessed. Once, the information sources are breaking down and reported, the Extract, Transform and Load (ETL)software is used to bring information from numerous sources and load into a staging area. From that point onward, the data purification process happens where information is isolated into clean and error table. The information in error table is then recleaned and changed into the perfect information. From here, information is stacked into a dimensional which is not normalized. The dimensional modelling is finished utilizing the star schema. The Kimball design approach uses the denormalized form for building entity structure. It is also based on conformed facts i.e., data marts which are separately implemented are grouped together with a robust architecture.

### Advantages of Bill Immon's Method:

The Inmon architecture offers the following advantages:

- The initial setup and execution is faster as there is no normalization involved.
- It simplifies querying and analysis as the data operators can be easily interpreted because of its denormalized structure.
- It takes less space in the database which makes system management simpler.
- A smaller team of designers and planners is sufficient for data warehouse management.
- It provides multi-dimensional structure and helps generate reliable insights.

### Disadvantages of Bill Immon's Method:

The Inmon architecture offers the following disadvantages:

In Kimball design, data isn't entirely integrated before reporting.

- As redundant data is added to database tables, data irregularities are most likely to occur.
- In the Kimball DW approach, the data warehouse model may be difficult to alter with any change in business needs.
- The model is business process-oriented so it won't focus on the other areas of the enterprise.

### **Assignment Portfolio**

### Data Warehouse design for LBU business:

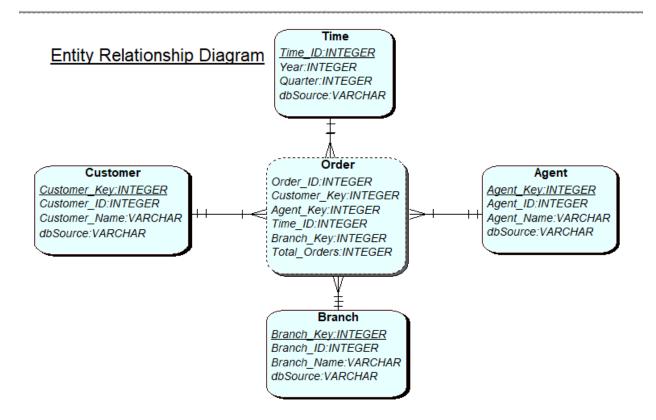
> Design the star schema for the DW to be implemented.

### **LBU Business:**

### Questions:

- 1. Design the star schema for the DW to be implemented....
- 2. Define, fact table, dimension(s), attributes, keys and measures
- 3. Number of orders made in UK first quarter of the year, in comparison with last year?
- 4. Who is our best customer, in first quarter of this year?
- 5. Total number of orders made in first quarter of the year, in comparison with last year for each branch?
- 6. Who is our best Agent, in the first quarter of this year?

Solution 1:



### Solution 2:

**Fact:** Order Table (<u>Order\_ID</u>, Customer\_Key, Agent\_Key, Time\_ID, Branch\_Key, Total\_Orders)

Dimensions: Customer (<u>Customer Key</u>, Customer\_ID, Customer\_Name, dbSource)

Agent (<u>Agent Key</u>, Agent\_ID, Agent\_Name, dbSource)

Branch (<u>Branch Key</u>, Branch\_ID, Branch\_Name, dbSource)

Time (Time ID, Year, Quarter, dbSource)

Measures: Total\_Orders

### **Solution 3:**

**SELECT** t.Quarter, SUM (Total\_Orders) **FROM** Order o, Time t

```
WHERE o.Time_ID = t.Time_ID
          AND t.Quarter = "Q1"
          AND TO_CHAR(SYSDATE, 'YYYY') = t.Year
          GROUP BY t.Quarter
          UNION
          SELECT t.Quarter, SUM (Total Orders)
          FROM Order o, Time t
          WHERE o.Time ID = t.Time ID
          AND TO_CHAR(SYSDATE, 'YYYY') -1 = t.Year
          GROUP BY t.Quarter;
Solution 4:
          SELECT SUM (Total_Orders), Customer_Key
          FROM
          SELECT SUM (Total Orders), Customer Key
          RANK OVER (ORDER BY SUM (Total Orders) DESC) AS Rank
          FROM Order o, Time t,
          WHERE o.Time_ID = t.Time_ID
          AND t.Quarter = "Q1"
          AND TO_CHAR(SYSDATE, 'YYYY') = t.Year
          WHERE Rank <=1;
Solution 5:
          SELECT t. Quarter, SUM (Total Orders), Branch Key
```

FROM Order o, Time t

```
WHERE o.Time_ID = t.Time_ID

AND t.Quarter = "Q1"

AND TO_CHAR(SYSDATE, 'YYYY') = t.Year

GROUP BY t.Quarter, Branch_Key

UNION

SELECT t.Quarter, SUM (Total_Orders), Branch_Key

FROM Order o, Time t

WHERE o.Time_ID = t.Time_ID

AND TO_CHAR(SYSDATE, 'YYYY') -1 = t.Year

GROUP BY t.Quarter, Branch_Key;
```

### **Solution 6:**

```
FROM
(

SELECT SUM (Total_Orders), Agent_Key

RANK OVER (ORDER BY SUM (Total_Orders) DESC) AS Rank

FROM Order o, Time t,

WHERE o.Time_ID = t.Time_ID

AND t.Quarter = "Q1"

AND TO_CHAR(SYSDATE, 'YYYY') = t.Year
)

WHERE Rank <=1;
```

# Exercise: Data Warehouse design for a wholesale furniture company

# Questions:

- 1. Identify facts, dimensions and measures
- 2. For each fact:
  - a) Produce the attribute tree and fact schema
  - b) Design the star or snowflake schema and write the following SQL queries:
    - i. Find the quantity, the total income and discount with respect to each city, type of furniture and the month
    - ii. Find the average quantity, income and discount with respect to each country, furniture material and year
    - iii. Determine the 5 most sold furniture during the May month

# Solutions:

### Question 1:

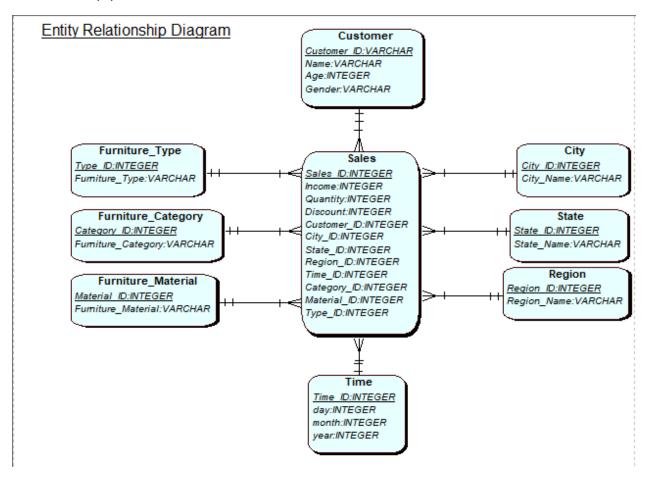
**FACT Sales** 

MEASURES Quantity, Income, Discount

DIMENSIONS Furniture (Type, Category, Material) Customer (Age, Sex, City  $\rightarrow$  Region  $\rightarrow$  State) Time (Day  $\rightarrow$  Month  $\rightarrow$  Year)

# Question 2 (a):

# Question 2(b):



# Question 2(b) i:

SELECT SUM(Quantity), SUM(Income), SUM(Discount), City\_ID, Type\_ID, Time\_ID FROM Sales

GROUP BY City\_ID, Type\_ID, Time\_ID;

# Question 2(b) ii:

```
SELECT AVG(Quantity), AVG(Income), AVG(Discount), State_ID, Material_ID, Time_ID
FROM Sales
GROUP BY State_ID, Material_ID, Time_ID;

Question 2(b) iii:

SELECT Type_ID, SUM(Quantity) as Total
FROM (
SELECT Type_ID, SUM(Quantity) as Total,
RANK() OVER (ORDER BY SUM(Quantity) DESC) as rank
FROM Sales s, Time t
WHERE t.Month = "May"
)
WHERE rank <= 5;
```