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قسم هندسة البرمجيات  
ونظم المعلومات

# **Building Business Intelligence Solution for The Syrian Ministry of Interior (Traffic Violations)**

بناء نظام ذكاء أعمال لـ  
وزارة الداخلية السورية  
(المخالفات المرورية)

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# SUPERVISION CERTIFICATION

I certify that the preparation of this project entitled **Building Business Intelligence Solution for The Syrian Ministry of Interior (Traffic Violations)**, prepared by **Ahmad Yones, Ibrahim Hafez, and Oumama Nabaa** was made under my supervision at *Department of Software & Information System Eng. – Faculty of Computer & Informatics Engineering* in partial fulfillment of the Requirements for the Degree of *Bachelors of Software & Information System Engineering*.

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Dr. Mouhib Alnoukari  
Supervisor  
Sunday, February 6, 2022

# Abstract:

Due to the rapid growth of new technologies, the Business Intelligence (BI) market is growing as well. Adoption of Business Intelligence system has become one of the most important technological and organizational innovations in modern organizations, and cornerstone of decision-making processes.

This project consists of implementing a business intelligence solution to the Ministry of Interior of Syria data, which increases decisions accuracy by providing accurate knowledge and easy-to-read reports to decision-makers.

This is accomplished by building a data warehouse relying on the extracted data, then analysis and business intelligence tools are applied to acquire the maximum benefits.

## ملخص:

نظراً للنمو السريع للتقنيات الجديدة، فإن سوق ذكاء الأعمال مستمر بالنمو أيضاً، فأصبح اعتماد نظام ذكاء الأعمال أحد أهم الابتكارات التكنولوجية والتنظيمية في المنظمات الحديثة، وحجر الأساس في عمليات صنع القرار.

يتكون هذا المشروع من تطبيق حلول ذكاء الأعمال لبيانات وزارة الداخلية السورية، مما يزيد من دقة القرارات من خلال توفير معرفة دقيقة وتقارير سهلة القراءة لصناع القرار.

يتم تحقيق ذلك من خلال بناء مستودع بيانات يعتمد على البيانات المستخرجة من الوزارة، ثم يتم تطبيق أدوات التحليل وذكاء الأعمال للحصول على الفائدة القصوى.

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# **Chapter 1**

## **Theoretical Study**

## 1.1. Introduction:

The world is becoming more and more data-driven, with endless amounts of data available to work with. Big companies like Google and Microsoft use data to make decisions, but they're not the only ones.

The importance of data analytics in any sector is compounded, creating enormous quantities of knowledge that can provide useful insights into the field. In the last ten years, this has led to a surge in the data market. In order to gain decision-making insights, the compilation of data can be supplemented by its analysis. Data analytics help organizations and businesses gain insight into the enormous amount of knowledge they need for further production and growth.

*“Information is the oil of the 21st century, and analytics is the combustion engine.”* – Peter Sondergaard, senior vice president, Gartner Research.

Business intelligence (BI) software can help by combining online analytical processing (OLAP), location intelligence, enterprise reporting, and more. BI software offers enterprise businesses the opportunity to connect disparate data sources into one unified source, collate and structure the data, and offer an interface for end-users to extract reports and dashboards that can drive more informed business decisions.



## **1.2. The Importance of The Project:**

The importance of the research lies in the following point work on important and relatively recent concepts, which are business intelligence systems in decision support:

- 1.2.1. A massive Shortage of decision support systems in Syria in general and in the governmental sectors in particular.
- 1.2.2. Understanding Your Government Customer” report, as people continue to demand a higher level of accountability and transparency from their government.
- 1.2.3. implementing a system to improve the level of services provided to beneficiaries in the governmental sector.
- 1.2.4. The need to develop analytical systems in order to take advantages of the large amount of data present in this organization to reach sound management decisions.
- 1.2.5. Effectively managing the allocation and utilization of funds requires in-depth insight into spending patterns.

### **1.3. The Goal of The Project:**

This project ensures the importance of building a business intelligence system in the Ministry of Interior, providing accurate knowledge for decision makers. goals in a nutshell:

- 1.3.1. Monitor optimal use of public funds.
- 1.3.2. Study Big data.
- 1.3.3. Task and workforce management.

## 1.4. Concepts:

### 1.4.1. Business Intelligence:

Business is the act of doing something productive to serve someone's needs, and thus earn a living and make the world a better place. Business activities are recorded on paper or using electronic media, and then these records become data. There is more data from customers' responses and on the industry as a whole. All this data can be analyzed and mined using special tools and techniques to generate patterns and intelligence, which reflect how the business is functioning. These ideas can then be fed back into the business so that it can evolve to become more effective and efficient in serving customer needs. And the cycle continues.

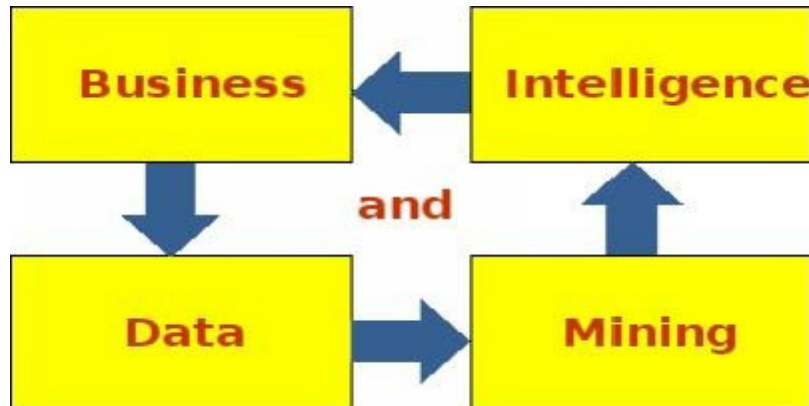


Figure 1.1 BI Cycle<sup>1</sup> (EMC Education Services (2014), *Data Science and Big Data Analytics*, p. 1)

BI is all about turning an organization's data into insights that can be used to inform business decisions. BI analysts will use BI tools, software or services to access and analyze datasets and translate their findings into reports, summaries, dashboards, graphs, charts or maps.

In recent years, the advent of modern data visualization and reporting tools has transformed the discipline, empowering businesses to use big data insights to identify, develop and create new business opportunities.

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<sup>1</sup> EMC Education Services (2014), *Data Science and Big Data Analytics*

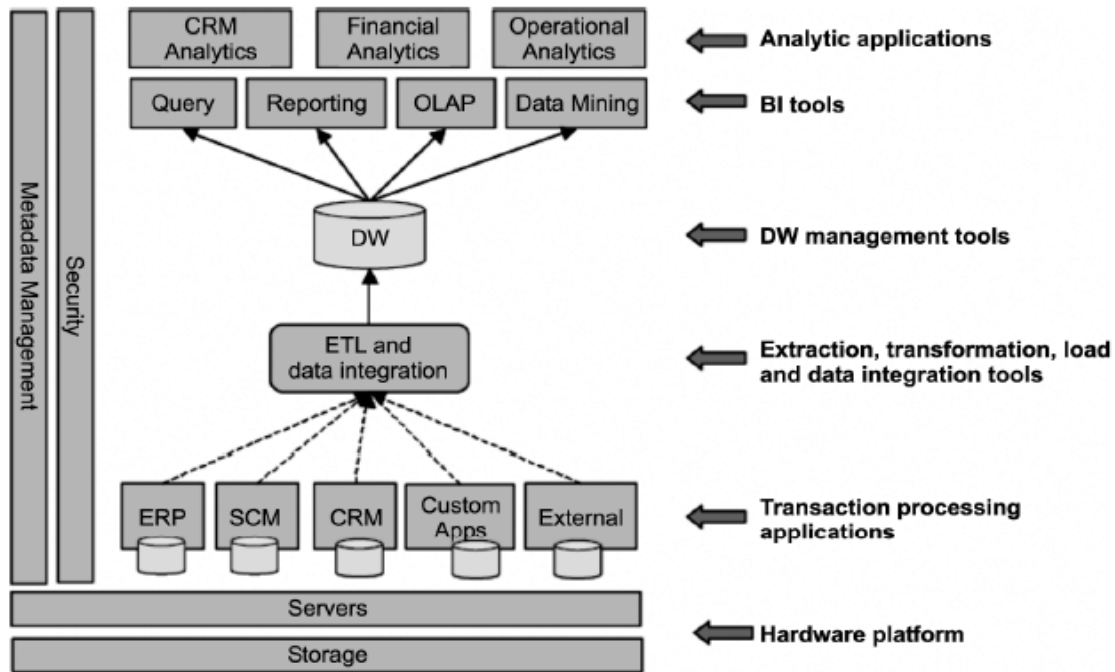


Figure 1.2 BI environment ([www.researchgate.net](http://www.researchgate.net))

The ultimate goal of BI initiatives is to drive better business decisions that enable organizations to increase revenue, improve operational efficiency and gain competitive advantages over business rivals. To achieve that goal, BI incorporates a combination of analytics, data management and reporting tools, plus various methodologies for managing and analyzing data.

BI data can include historical information and real-time data gathered from source systems as it's generated, enabling BI tools to support both strategic and tactical decision-making processes. Before it's used in BI applications, raw data from different source systems generally must be integrated, consolidated and cleansed using data integration and data quality management tools to ensure that BI teams and business users are analyzing accurate and consistent information.

### 1.4.2. Business Intelligence:

Business intelligence (BI) is all about converting large amounts of corporate data into useful information, thereby triggering some profitable business action with the help of knowledge acquired through BI analysis.

Implementing BI is a long process and it requires a lot of analysis and investment. A typical BI environment involves business models, data models, data sources, ETL, tools needed to transform and organize the data into useful information, target data warehouse, data marts, OLAP analysis and reporting tools.

Setting up a BI environment not only rely on tools, techniques and processes, it also requires skilled business people to carefully drive these in the right direction. Care should be taken in understanding the business requirements, setting up the targets, analyzing and defining the various processes associated with these, determining what kind of data needed to be analyzed, determining the source and target for that data, defining how to integrate that data for BI analysis and determining and gathering the tools to achieve this goal.

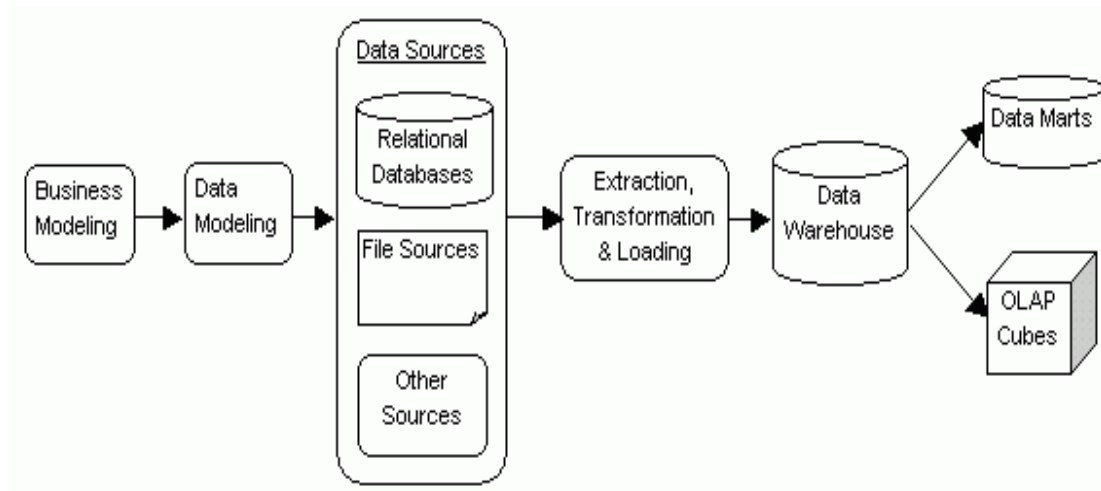


Figure 1.3 Simpler BI environment (bimentalist.com)

The importance of business intelligence continues to grow as businesses face an ever-increasing flow of raw data and the challenges of gaining insight from enormous volumes of information (big data). With the employment of business intelligence systems, businesses can

gain a comprehensive view of their organization's data and translate it into insights about their business processes, enabling improved and strategic business decisions. Business intelligence helps organizations analyze data with a historical context, optimize operations, track performance, accelerate and improve decision-making, identify and eliminate business problems and inefficiencies, identify market trends and patterns, drive new revenues and profitability, increase productivity and accelerate growth, analyze customer behavior, compare data with competitors, and ultimately gain a competitive advantage over rival businesses.

- 1- Faster analysis, intuitive dashboards.
- 2- Increased organizational efficiency.
- 3- Data-driven business decisions.
- 4- Improved customer experience.
- 5- Improved employee satisfaction.
- 6- Trusted and governed data.

Modern business intelligence systems prioritize self-service analysis, empowering businesses to gain insight into their market and improve performance with comprehensive data discovery tools, methods, processes, and platforms. Such business intelligence solutions include:

- 1- Ad hoc analytics: an analysis process designed to answer specific questions on the spot.
- 2- Online analytical processing (OLAP): a computing method that enables multi-dimensional analytical queries
- 3- Mobile BI: software that optimizes desktop business intelligence for mobile devices.

Real-time BI: a data analytics approach that delivers real-time information to users by feeding business transactions into a real-time data warehouse Operational BI: a data analysis approach that utilizes real-time business analytics to automatically integrate

real-time data into operational system for immediate use.

Software-as-a-service BI (SaaS BI): a cloud-hosted, subscription-based delivery model for business intelligence software solutions

Open-source BI (OSBI): business intelligence software solutions that do not require purchasing a software license

Collaborative BI: the merging of business intelligence software with collaboration tools in order to streamline the sharing process

Location intelligence (LI): software that is designed to relate geographic contexts to business data visualization software: facilitates the detection of patterns and correlations by providing visual context.

### 1.4.3. Data Warehouse (DW):

A Data Warehousing (DW) is process for collecting and managing data from varied sources to provide meaningful business insights. A Data warehouse is typically used to connect and analyze business data from heterogeneous sources. The data warehouse is the core of the BI system which is built for data analysis and reporting. It is a blend of technologies and components which aids the strategic use of data. It is electronic storage of a large amount of information by a business which is designed for query and analysis instead of transaction processing. It is a process of transforming data into information and making it available to users in a timely manner to make a difference.

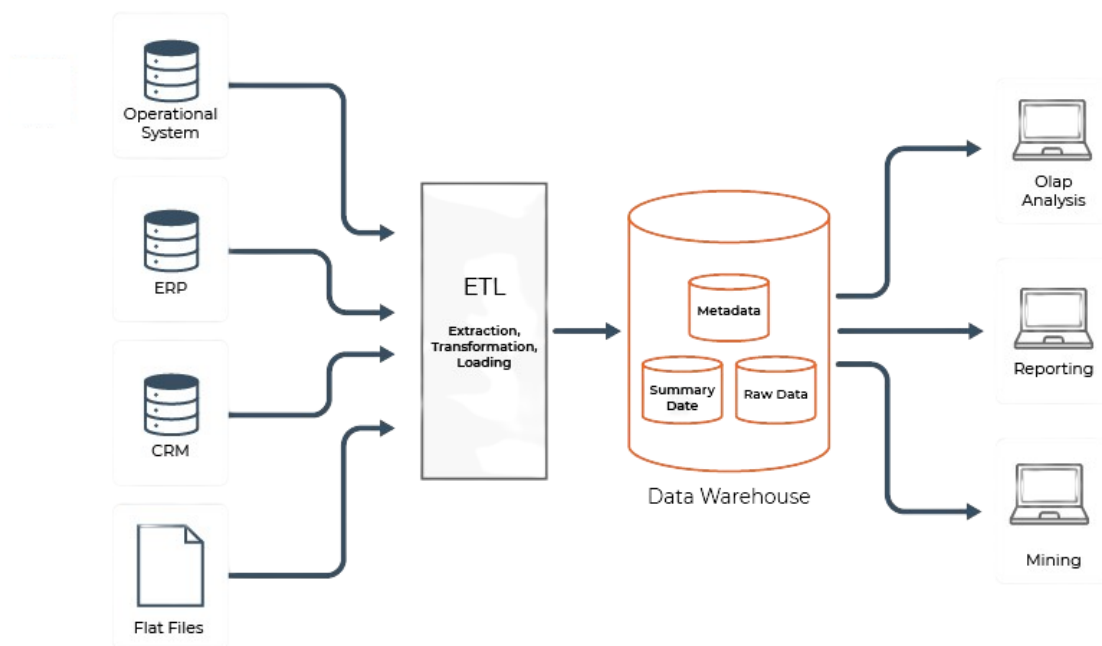


Figure 1.4 Data Warehouse Architecture ([www.datawarehouse4u.info](http://www.datawarehouse4u.info))



#### 1.4.4. Extract, Transform, and Load:

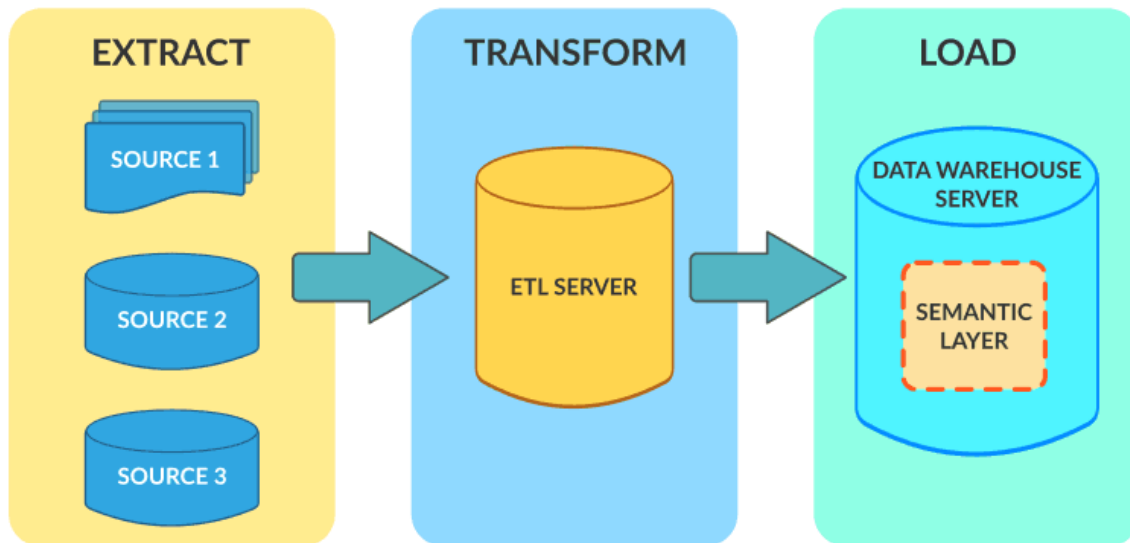


Figure 1.5 ETL tool (thecustomizewindows.com)

Thus, an ETL tool extracts data from disparate sources, transforms it to make it compatible with the destination system, and then loads it into the destination system. This destination could be a data warehouse, data lake, database, or any other application system.

A properly designed ETL system extracts data from the source systems, enforces data quality and consistency standards, conforms data so that separate sources can be used together, and finally delivers data in a presentation-ready format so that application developers can build applications and end users can make decisions.

Since the data extraction takes time, it is common to execute the three phases in pipeline. While the data is being extracted, another transformation process executes while processing the data already received and prepares it for loading while the data loading begins without waiting for the completion of the previous phases.

#### 1.4.4.1. Extract:

In this step of ETL architecture, data is extracted from the source system into the staging area. Transformations if any are done in staging area so that performance of source system is not degraded. Also, if corrupted data is copied directly from the source into Data warehouse database, rollback will be a challenge. Staging area gives an opportunity to validate extracted data before it moves into the Data warehouse.

Data warehouse needs to integrate systems that have different DBMS, Hardware, Operating Systems and Communication Protocols. Sources could include legacy applications like Mainframes, customized applications, point of contact devices like ATM, Call switches, text files, spreadsheets, ERP, data from vendors, partners amongst others.

Hence one needs a logical data map before data is extracted and loaded physically. This data map describes the relationship between sources and target data. Three Data Extraction Methods:

1. Full Extraction.
2. Partial Extraction- without update notification.
3. Partial Extraction- with update notification.

Irrespective of the method used, extraction should not affect performance and response time of the source systems. These source systems are live production databases. Any slow down or locking could affect company's bottom line.

#### 1.4.4.2. Transform:

Data extracted from source server is raw and not usable in its original form. Therefore, it needs to be cleansed, mapped and transformed. In fact, this is the key step where ETL process adds value and changes data such that insightful BI reports can be generated.

It is one of the important ETL concepts where you apply a set of functions on extracted data. Data that does not require any transformation is called as direct move or pass-through data.

In transformation step, you can perform customized operations on data. For instance, if the

user wants sum-of-sales revenue which is not in the database. Or if the first name and the last name in a table is in different columns. It is possible to concatenate them before loading.

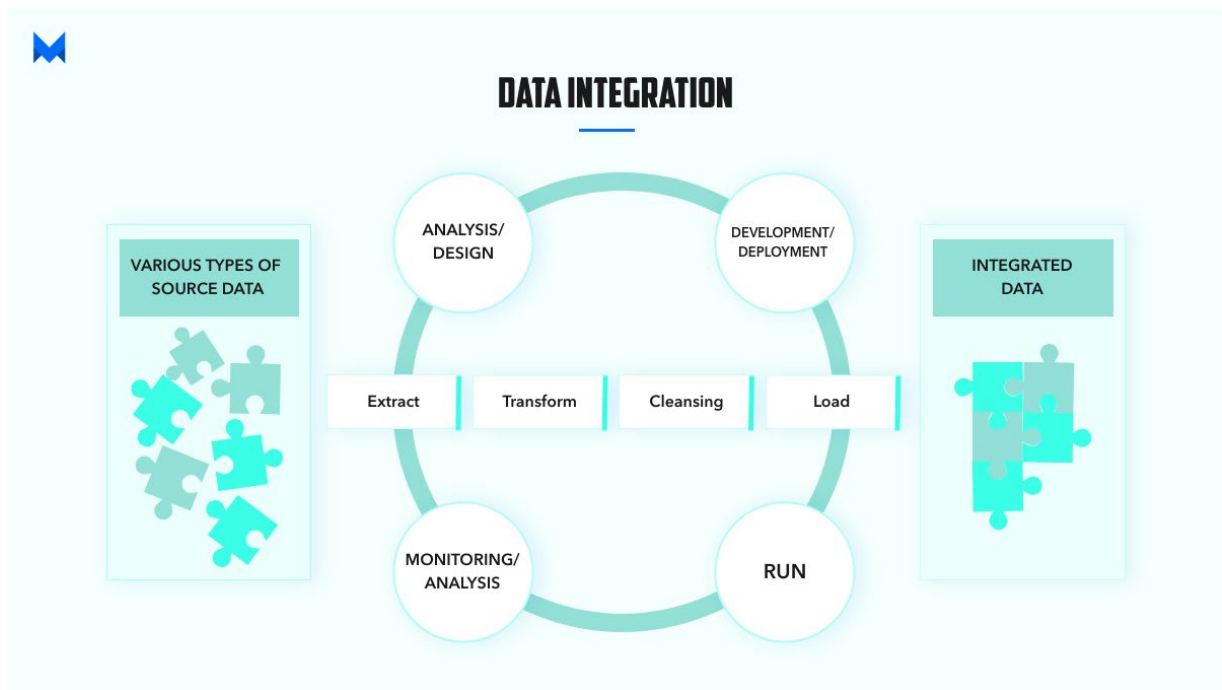


Figure 1.6 Data Integration Cycle ([www.mobcoder.com](http://www.mobcoder.com))

Data integration is a general industry term referring to the requirement to combine data from multiple sources for a single unified view to process it further into insights. In separate words, it is a practice to consolidate data from multiple sources into a single dataset or unified view with the aim of providing users with consistent access across all data types.

**Here are the challenges and Data integration criteria:** There are three major data integration criteria to build a data warehouse are:

**Freshness:** The process of data integration architecture is executed at certain periods. In this, data freshness refers to the delay when data is tweaked on a source system to when the change is executed in the data warehouse.

**Granularity:** Data from the source system is channelized or aggregated in the very process of data integration. Here the data granularity refers to the level of detail with data stored in the data warehouse.

**Matching:** In the data integration process, data comes from disparate sources and needs to be specified in a specific category. Here matching defines how data from multiple sources should be matched can be complex.

#### 1.4.4.3. Load:

Loading is the ultimate step in the ETL process. In this step, the extracted data and the transformed data are loaded into the target database. To make the data load efficient, it is necessary to index the database and disable the constraints before loading the data. All three steps in the ETL process can be run parallel. Data extraction takes time and therefore the second phase of the transformation process is executed simultaneously. This prepared the data for the third stage of loading. As soon as some data is ready, it is loaded without waiting for the previous steps to be completed.

The loading process is the physical movement of the data from the computer systems storing the source database(s) to that which will store the data warehouse database. The entire process of transferring data to a data warehouse repository is referred to in the following ways:

- 1- Initial Load: For the very first time loading all the data warehouse tables.
- 2- Incremental Load: Periodically applying ongoing changes as per the requirement. After the data is loaded into the data warehouse database, verify the referential integrity between the dimensions and the fact tables to ensure that all records belong to the appropriate records in the other tables. The DBA must verify that each record in the fact table is related to one record in each dimension table that will be used in combination with that fact table.
- 3- Full Refresh: Deleting the contents of a table and reloading it with fresh data.

**Refresh versus Update** After the initial load, the data warehouse needs to be maintained and updated and this can be done by the following two methods:

Update-application of incremental changes in the data sources.

Refresh-complete reloads at specified intervals.

**Data Loading:** Data is physically moved to the data warehouse. The loading takes place

within a “load window. The tendency is close to real-time updates for data warehouses as warehouses are growing used for operational applications.

**Loading the Dimension Tables:** Procedure for maintaining the dimension tables includes two functions, initial loading of the tables and thereafter applying the changes on an ongoing basis. System generated keys are used in a data warehouse. The records in the source system have their own keys. Therefore, before an initial load or an ongoing load, the production keys must be set to system-generated keys in the data warehouse, another issue is related to the application of Type 1, Type 2, and Type 3 changes to the data warehouse. Fig. shows how to handle it.

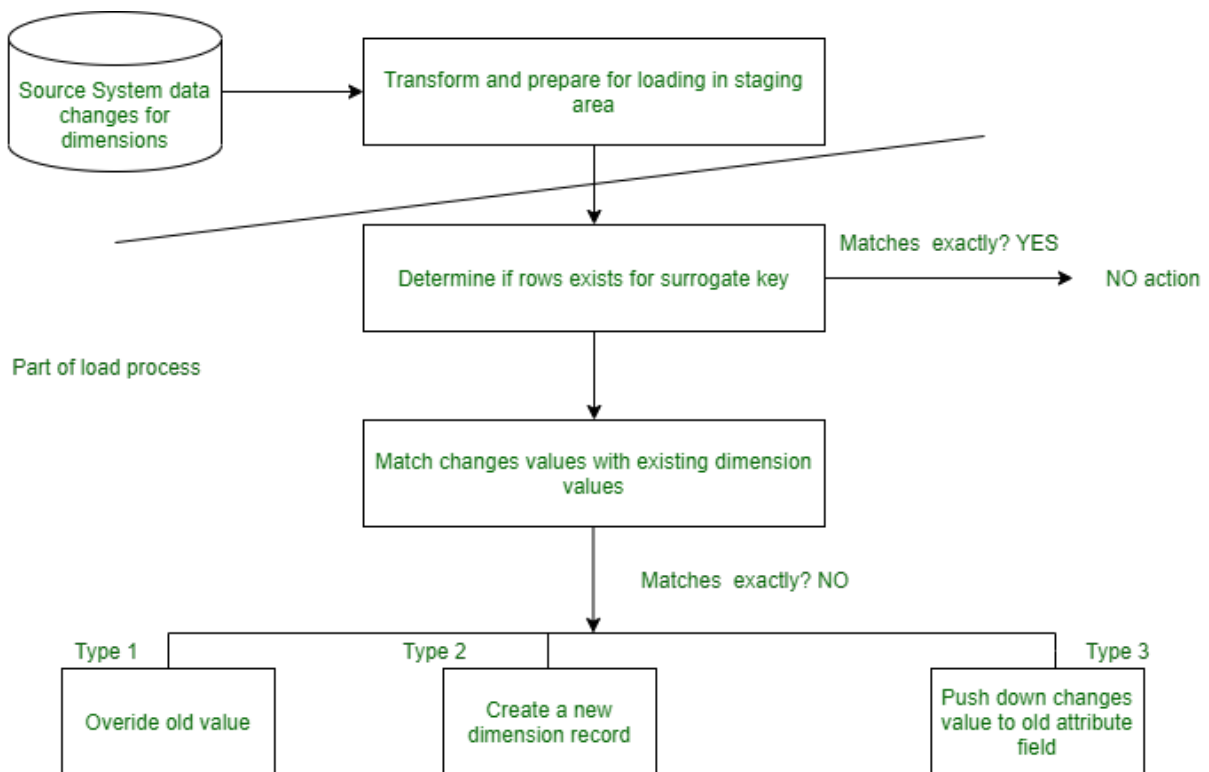


Figure 1.7 Loading changes to a dimension table ([www.geeksforgeeks.com](http://www.geeksforgeeks.com))

### **1.4.5. Data Mining:**

#### **1.4.5.1. Introduction:**

In general terms, “Mining” is the process of extraction of some valuable material from the earth e.g., coal mining, diamond mining, etc. In the context of computer science, “Data Mining” can be referred to as knowledge mining from data, knowledge extraction, data/pattern analysis, data archaeology, and data dredging. It is basically the process carried out for the extraction of useful information from a bulk of data or data warehouses. One can see that the term itself is a little confusing. In the case of coal or diamond mining, the result of the extraction process is coal or diamond. But in the case of Data Mining, the result of the extraction process is not data!! Instead, data mining results are the patterns and knowledge that we gain at the end of the extraction process. In that sense, we can think of Data Mining as a step in the process of Knowledge Discovery or Knowledge Extraction.

Gregory Piatetsky-Shapiro coined the term “Knowledge Discovery in Databases” in 1989. However, the term ‘data mining’ became more popular in the business and press communities. Currently, Data Mining and Knowledge Discovery are used interchangeably. Nowadays, data mining is used in almost all places where a large amount of data is stored and processed. For example, banks typically use ‘data mining’ to find out their prospective customers who could be interested in credit cards, personal loans, or insurance as well. Since banks have the transaction details and detailed profiles of their customers, they analyze all this data and try to find out patterns that help them predict that certain customers could be interested in personal loans, etc.

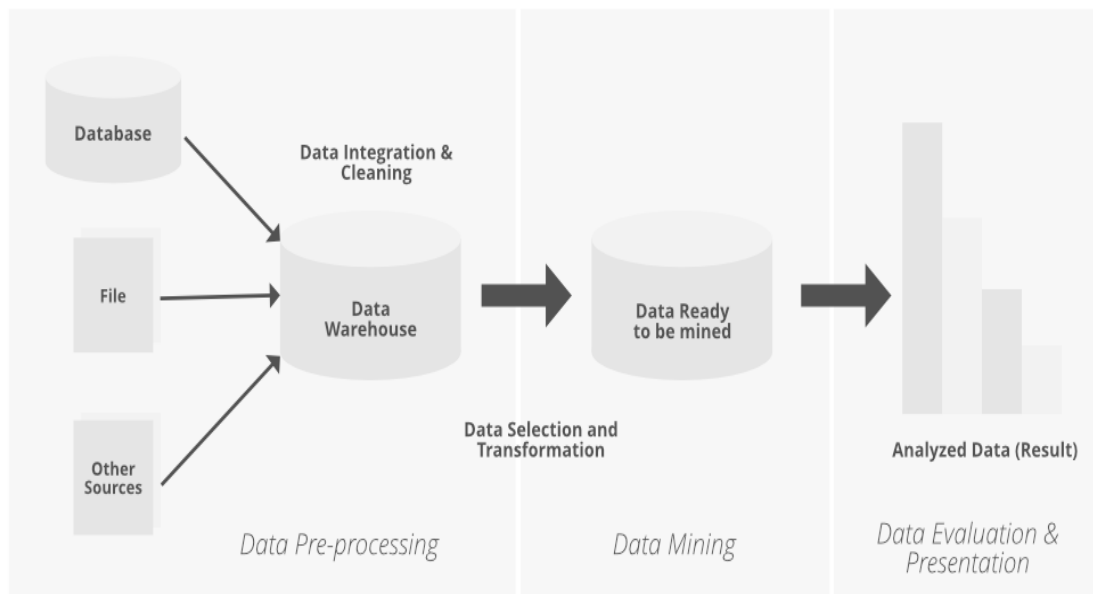


Figure 1.8 Level of Analytics

#### **1.4.5.2. Methods and Techniques:**

Data may be mined to help make more efficient decisions in the future. Or it may be used to explore the data to find interesting associative patterns. The right technique depends upon the kind of problem being solved. (Association, Classification, Predication, Clustering, Regression, ...).

##### **a. Association:**

Association analysis is the finding of association rules showing attribute-value conditions that occur frequently together in a given set of data. Association analysis is widely used for a market basket or transaction data analysis. Association rule mining is a significant and exceptionally dynamic area of data mining research. One method of association-based classification, called associative classification, consists of two steps. In the main step, association instructions are generated using a modified version of the standard association rule mining algorithm known as Apriori. The second step constructs a classifier based on the association rules discovered.

##### **b. Classification:**

Classification is the processing of finding a set of models (or functions) that describe and distinguish data classes or concepts, for the purpose of being able to use the model to predict the class of objects whose class label is unknown. The determined model depends on the investigation of a set of training data information (i.e. data objects whose class label is known). The derived model may be represented in various forms, such as classification (if – then) rules, decision trees, and neural networks.

##### **c. Prediction:**



Data Prediction is a two-step process, similar to that of data classification. Although, for prediction, we do not utilize the phrasing of “Class label attribute” because the attribute for which values are being predicted is consistently valued(ordered) instead of categorical (discrete-esteemed and unordered). The attribute can be referred to simply as the predicted attribute. Prediction can be viewed as the construction and use of a model to assess the class of an unlabeled object, or to assess the value or value ranges of an attribute that a given object is likely to have.

#### **d. Clustering:**

Unlike classification and prediction, which analyze class-labeled data objects or attributes, clustering analyzes data objects without consulting an identified class label. In general, the class labels do not exist in the training data simply because they are not known to begin with. Clustering can be used to generate these labels. The objects are clustered based on the principle of maximizing the intra-class similarity and minimizing the interclass similarity. That is, clusters of objects are created so that objects inside a cluster have high similarity in contrast with each other, but are different objects in other clusters. Each Cluster that is generated can be seen as a class of objects, from which rules can be inferred. Clustering can also facilitate classification formation, that is, the organization of observations into a hierarchy of classes that group similar events together.

#### **e. Regression**

Regression can be defined as a statistical modeling method in which previously obtained data is used to predicting a continuous quantity for new observations. This classifier is also known as the Continuous Value Classifier. There are two types of regression models: Linear regression and multiple linear regression models.

# **Chapter 2**

## **Design and Analytical Study (Implementation)**

## 2.1. Introduction:

Business intelligence (BI) comprises the strategies and technologies used by enterprises for the data analysis of business information. BI systems leverage software and services beginning with collecting data, cleansing data, integration of data, and then using data visualization to transform data into actionable insights that might be helpful while decision making process. Implementing a business intelligence system requires careful planning to assure that it meets expectations. These are the basic steps:

- 1- Identify End-User Requirements
- 2- Identify the Data Sources (Data sources: which are reports in excel forms.)
- 3- ETL (Extract, Transform, and Load): in this area we managed to gather them, perform cleaning operations, and then visualize the data in a better form.
- 4- Design the Data Model
- 5- The Data Warehouse: Which contains all the relevant data, and can be described as an enterprise database.
- 6- The OLAP environment.
- 7- The BI application.

## **2.2. Identify End-User Requirements:**

- ✓ What id the relation between number of violations and the quarters of year
- ✓ What is the relation between number of violations and License City?
- ✓ What is the relation between city of violation and city of license?
- ✓ How many numbers of violation paid after 1 month?
- ✓ How many numbers of violations paid in the same year?
- ✓ How many SP we get from a specific biller?
- ✓ What is the avg of discounted points?

## 2.3. Identify the Data sources:

We had the data directly from the Ministry of Interior provided by our supervisor Dr. Mouhib Alnoukari. It's an Excel file with 402540 records, 17 columns, and a size of **45.7MB**

No	Date	ViolationCity	ViolationType	IsPresented	TAB#	CarType	CarCity	VAL	END_VL	POIN	END_POIN	END_DATE	ECity	Billor
430166	6/5/2021	Tartus	دفع المخالفة	1	102430	عامة	Tartus	4,000	2,000	4	4	6/6/21 17:59	Tartus	NULL
430166	6/5/2021	Tartus	دفع المخالفة	1	102430	عامة	Tartus	7,000	3,500	6	6	6/6/21 17:59	Tartus	NULL
176791	8/25/2021	Tartus	دفع المخالفة	1	705531	عامة	Tartus	7,000	3,500	6	6	9/3/21 1:16	Tartus	NULL
176791	8/25/2021	Tartus	دفع المخالفة	1	705531	عامة	Tartus	500	250	0	0	9/3/21 1:16	Tartus	NULL
478416	9/21/2021	Aleppo	دفع المخالفة	1	715180	خاصة	Tartus	7,000	3,500	6	6	9/23/21 13:43	Aleppo	NULL
133253	9/27/2021	Tartus	دفع المخالفة	1	715644	خاصة	Tartus	2,000	1,000	0	0	9/28/21 16:42	Tartus	NULL
132259	9/15/2021	Tartus	دفع المخالفة	1	715825	خاصة	Tartus	4,000	2,000	4	4	9/21/21 14:50	Tartus	NULL
439758	6/19/2021	Tartus	دفع المخالفة	1	302822	خاصة	Tartus	2,000	1,000	0	0	6/22/21 13:41	Tartus	NULL
426190	8/29/2021	Tartus	دفع المخالفة	0	309614	خاصة	Tartus	5,000	2,500	4	4	9/26/21 12:34	Homs	برع مرور حمص
426190	8/29/2021	Tartus	دفع المخالفة	0	309614	خاصة	Tartus	7,000	3,500	6	6	9/26/21 12:34	Homs	برع مرور حمص
900902	8/18/2021	Tartus	دفع المخالفة	0	309614	خاصة	Tartus	5,000	2,500	4	4	9/26/21 12:34	Homs	برع مرور حمص
900902	8/18/2021	Tartus	دفع المخالفة	0	309614	خاصة	Tartus	7,000	3,500	6	6	9/26/21 12:34	Homs	برع مرور حمص
425893	6/3/2021	Tartus	دفع المخالفة	1	309614	خاصة	Tartus	4,000	2,000	4	4	6/7/21 14:01	Tartus	NULL
133075	9/20/2021	Tartus	دفع المخالفة	1	309614	خاصة	Tartus	4,000	2,000	4	4	9/29/21 15:08	Tartus	NULL
133075	9/20/2021	Tartus	دفع المخالفة	1	309614	خاصة	Tartus	4,000	2,000	4	4	9/29/21 15:08	Tartus	NULL
542881	5/4/2021	Damascus	دفع المخالفة	0	309194	خاصة	Tartus	7,000	3,500	6	6	8/11/21 10:37	Latakia	برع مرور اللاذقية
623797	6/24/2021	Hama	دفع المخالفة	1	319006	خاصة	Tartus	4,000	2,000	4	4	6/30/21 13:21	Hama	NULL
885934	10/5/2021	Homs	دفع المخالفة	1	486067	خاصة	Tartus	4,000	2,000	4	4	10/10/21 14:14	Homs	NULL
436882	5/7/2021	Tartus	دفع المخالفة	1	327137	خاصة	Tartus	2,000	1,000	2	2	5/9/21 21:03	Tartus	NULL
437255	8/31/2021	Tartus	دفع المخالفة	1	308090	خاصة	Tartus	7,000	3,500	6	6	9/2/21 9:47	Tartus	NULL
436342	5/21/2021	Tartus	دفع المخالفة	1	308557	خاصة	Tartus	500	250	0	0	5/24/21 10:55	Tartus	NULL
478446	6/28/2021	Latakia	دفع المخالفة	1	327348	خاصة	Tartus	3,000	1,000	0	0	7/5/21 8:08	Latakia	NULL

## 2.4. ETL:

Our biggest challenge was the data cleansing phase; we got into so many problems due to the low-quality data and the massive number of missing values, failed entries, and data ambiguity (We couldn't find someone to explain for us some attributes of the provided data).

We used Pandas and NumPy libraries for generating data quality reports, excel and power BI for dropping columns and filling null values.

### Data cleansing phase:

- Dropping records that consist of duplicated entries, wrong entries, conflicts, and ambiguous meanings. Duplicated Entries

No	Date	Location	ViolationType	IsPresented	TABNI	CarType	CarCh	VAL	END_V	POINT	NO_POI	END_DATE	ECITY	Biller	license	licenset	licenset	e
18022	16050	2021-03-03 00:00:00	Hama	الاحداث مكررة	1.460201	عامة	Raaga	7000	3500	6	6	2021-05-09 13:04:32	Hama	329602	Hama	خاصة ب		
49892	43782	2021-08-10 00:00:00	Hama	الاحداث مكررة	1.454003	خاصة	Tartus	4000	2000	4	4	2021-08-19 12:02:34	Tartus	119721	Tartus	خاصة ب		
51566	51684	2021-06-25 00:00:00	Aleppo	الاحداث مكررة	1.635134	خاصة	Aleppo	7000	3500	6	6	2021-06-27 09:04:13	Aleppo	250146	Al-Hasaka	خاصة ب		
66531	66679	2021-05-07 00:00:00	Hama	الاحداث مكررة	1.735470	عامة	Hama	2000	1000	0	0	2021-05-08 13:10:54	Hama	1836	Hama	عامة 1د		
66532	66680	2021-05-07 00:00:00	Hama	الاحداث مكررة	1.735470	عامة	Hama	4000	2000	4	4	2021-05-08 13:11:00	Hama	1836	Hama	عامة 1د		
66879	67028	2021-06-29 00:00:00	Damascus	الاحداث مكررة	0.879648	خاصة	Damascus	4000	2000	4	4	2021-07-24 08:41:57	Damascus					
67007	67157	2021-07-22 00:00:00	Damascus	الاحداث مكررة	0.610430	عامة	Damascus	2000	1000	0	0	2021-09-06 09:24:52	Damascus					
68270	68425	2021-07-27 00:00:00	Damascus	الاحداث مكررة	0.638043	خاصة	Damascus	2000	1000	0	0	2021-08-15 14:05:21	Damascus					
70584	70741	2021-06-15 00:00:00	Damascus	الاحداث مكررة	0.882501	خاصة	Damascus	2000	1000	0	0	2021-07-11 14:19:58	Damascus					
70612	70769	2021-05-11 00:00:00	Damascus	الاحداث مكررة	0.702192	خاصة	Damascus	7000	3500	6	6	2021-09-23 09:44:47	Damascus					
70685	70822	2021-07-24 00:00:00	Damascus	الاحداث مكررة	0.891676	خاصة	Damascus	2000	1000	0	0	2021-09-30 15:37:43	Damascus					
70887	71144	2021-07-25 00:00:00	Damascus	الاحداث مكررة	0.977947	خاصة	Damascus	2000	1000	0	0	2021-08-30 06:56:14	Damascus					
74003	74162	2021-07-25 00:00:00	Damascus	الاحداث مكررة	0.420613	عامة	Damascus	5000	2500	4	4	2021-09-01 09:53:39	Damascus					
76532	76698	2021-06-10 00:00:00	Damascus	الاحداث مكررة	0.726606	خاصة	Damascus	7000	3500	6	6	2021-07-05 11:28:15	Damascus					
76618	76784	2021-07-24 00:00:00	Damascus	الاحداث مكررة	0.884056	خاصة	Damascus	4000	2000	4	4	2021-09-09 11:25:22	Damascus					
77833	78001	2021-06-06 00:00:00	Latakia	الاحداث مكررة	0.887697	خاصة	Hama	4000	2000	4	4	2021-08-14 13:45:59	Hama					
78081	78229	2021-09-16 00:00:00	As-Suway	الاحداث مكررة	0.717074	خاصة	Damascus	500	250	0	0	2021-10-11 08:28:48	Damascus					
78717	78885	2021-06-08 00:00:00	Damascus	الاحداث مكررة	0.442000	خاصة	Damascus	2000	1000	0	0	2021-09-01 13:26:47	Damascus					
81872	82048	2021-06-01 00:00:00	Damascus	الاحداث مكررة	0.132098	خاصة	Damascus	2000	1000	0	0	2021-06-13 12:40:52	Damascus					
82971	83150	2021-07-25 00:00:00	Damascus	الاحداث مكررة	0.569390	خاصة	Damascus	2000	1000	0	0	2021-10-11 12:23:12	Damascus					
84696	84878	2021-07-27 00:00:00	Damascus	الاحداث مكررة	0.971998	خاصة	Damascus	7000	3500	6	6	2021-08-22 10:25:48	Damascus					
86026	86211	2021-08-19 00:00:00	Damascus	الاحداث مكررة	0.737139	خاصة	Damascus	2000	1000	0	0	2021-09-07 13:54:53	Damascus					
86070	86255	2021-07-26 00:00:00	Damascus	الاحداث مكررة	0.135173	خاصة	Damascus	2000	1000	0	0	2021-10-02 13:15:29	Damascus					
86214	86399	2021-09-29 00:00:00	Damascus	الاحداث مكررة	0.337291	خاصة	Damascus	2000	1000	0	0	2021-10-05 13:06:56	Damascus					
88772	88918	2021-07-25 00:00:00	Damascus	الاحداث مكررة	0.726655	خاصة	Damascus	2000	1000	0	0	2021-08-15 16:56:42	Damascus					
88911	89098	2021-07-08 00:00:00	Damascus	الاحداث مكررة	0.339997	خاصة	Damascus	2000	1000	0	0	2021-08-20 12:15:22	Damascus					
92222	92418	2021-09-21 00:00:00	Damascus	الاحداث مكررة	0.104944	خاصة	Damascus	2000	1000	0	0	2021-10-11 08:33:56	Damascus					
93699	93895	2021-07-03 00:00:00	Damascus	الاحداث مكررة	0.831272	خاصة	Damascus	2000	1000	0	0	2021-08-01 12:01:40	Damascus					
94958	95158	2021-07-28 00:00:00	Damascus	الاحداث مكررة	0.502489	خاصة	Damascus	5000	2500	4	4	2021-09-05 11:12:22	Damascus					
95508	95709	2021-09-18 00:00:00	Damascus	الاحداث مكررة	0.735543	خاصة	Damascus	5000	2500	4	4	2021-09-19 08:13:52	Damascus					
99078	99288	2021-08-01 00:00:00	Damascus	الاحداث مكررة	0.949424	خاصة	Damascus	2000	1000	0	0	2021-09-12 13:09:43	Damascus					
99747	99960	2021-07-23 00:00:00	Damascus	الاحداث مكررة	0.107850	خاصة	Damascus	5000	2500	4	4	2021-08-11 10:57:54	Damascus					
101313	101530	2021-06-06 00:00:00	Damascus	الاحداث مكررة	0.946576	خاصة	Damascus	2000	1000	0	0	2021-09-20 12:51:56	Damascus					
102253	102469	2021-07-27 00:00:00	Damascus	الاحداث مكررة	0.946107	خاصة	Damascus	7000	3500	6	6	2021-08-12 08:56:32	Damascus					
104832	104853	2021-07-26 00:00:00	Damascus	الاحداث مكررة	0.948059	خاصة	Damascus	2000	1000	0	0	2021-08-29 09:19:40	Damascus					
104940	104962	2021-09-17 00:00:00	Rif Dimas	الاحداث مكررة	0.947024	خاصة	Damascus	2000	1000	0	0	2021-10-10 11:46:58	Damascus					
105463	105688	2021-07-24 00:00:00	Damascus	الاحداث مكررة	0.503843	خاصة	Damascus	2000	1000	0	0	2021-10-06 10:12:17	Damascus					

- Wrong Entities

No	Date	Location	ViolationType	IsPresented	TABNI	CarType	CarCh	VAL	END_V	POINT	NO_POI	END_DATE	ECITY	Biller	license	licenset	licenset	e
12206	12205	2021-08-10 00:00:00	Hama	خطأ	1.75505	خاصة	Rif Dimas	4000	2000	4	4	2021-08-22 13:15:45	Hama	210143	Hama	خاصة ب		
37697	37696	2021-05-02 00:00:00	Hama	خطأ	1.603629	خاصة	Hama	1000	500	0	0	2021-05-08 14:27:38	Hama	248298	Hama	خاصة ب		
46316	46315	2021-05-18 00:00:00	Hama	خطأ	1.532639	خاصة	Idlib	4000	2000	4	4	2021-05-08 11:39:26	Hama	35272	Hama	خاصة ب		
47975	47974	2021-10-02 00:00:00	Damascus	خطأ	1.525977	خاصة	Damascus	4000	2000	4	4	2021-10-02 23:19:20	Damascus	826074	Damascus	خاصة ب		
59869	59868	2021-09-27 00:00:00	Hama	خطأ	1.795497	خاصة	Hama	7000	3500	6	6	2021-10-03 11:04:21	Hama	148219	Hama	خاصة ب		
298614	298613	2021-07-01 00:00:00	Daraa	خطأ	0.573046	خاصة	Daraa	15000	7500	8	8	2021-07-11 12:52:54	Daraa					
298615	298614	2021-07-01 00:00:00	Daraa	خطأ	0.573046	خاصة	Daraa	7000	3500	6	6	2021-07-11 12:53:31	Daraa					
298616	298615	2021-07-01 00:00:00	Daraa	خطأ	0.573046	خاصة	Daraa	5000	2500	4	4	2021-07-11 12:53:16	Daraa					
342093	342094	2021-10-03 00:00:00	Hama	خطأ	1.521894	خاصة	Hama	4000	2000	4	4	2021-10-03 12:21:35	Hama		4400	Hama	عامة 1د	
342096	342095	2021-10-03 00:00:00	Hama	خطأ	1.521894	خاصة	Hama	1000	500	0	0	2021-10-03 12:21:24	Hama		4400	Hama	عامة 1د	
342124	342123	2021-05-02 00:00:00	Hama	خطأ	1.740490	خاصة	Hama	4000	2000	4	4	2021-05-08 14:28:12	Hama	26382	Hama	عامة صغيرة		
342125	342124	2021-05-02 00:00:00	Hama	خطأ	1.740490	خاصة	Hama	500	250	0	0	2021-05-08 14:28:23	Hama	26382	Hama	عامة صغيرة		
346217	346216	2021-09-08 00:00:00	Hama	خطأ	1.638239	خاصة	Hama	7000	3500	6	6	2021-09-09 12:54:16	Idlib	48751	Hama	عامة 2د		
389221	389220	2021-09-08 00:00:00	Hama	خطأ	1.733636	استخبارات	Hama	500	250	0	0	2021-09-13 13:34:36	Idlib	84528	Hama	عامة 2د		
401964	401963	2021-08-16 00:00:00	Hama	خطأ	1.808599	خاصة	Idlib	7000	3500	6	6	2021-08-17 08:57:05	Hama	11135	Idlib	عامة 2د		

- Conflicts

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	
1	id	date	city	type	preser	total_v	end_v	diff_va	point	end_poi	diff_poi	end_date	e_city	billr_na	tab_n	car_cit	car_tyt	lic_nc	lic_cit	lic_tyt	is_knov	
64255	64439	07/07/2021	Rif Dimashq	فيد الانتظار	1	500	250	250	0	3	-3		Rif Dimashq	..0000		Rif Dimashq	خاصة	461953	Rif Dimashq	خاصة ب		
66191	66379	07/29/2021	Rif Dimashq	فيد الانتظار	1	2000	1000	1000	0	4	-4		Rif Dimashq	293		Rif Dimashq	خاصة	573287	Rif Dimashq	خاصة ب		
210945	211597	05/08/2021	Latakia	دفع المخالفة	0	7000	7000	0	6	12	-6	#####	Latakia	Latakia Tr	017632		Latakia	عامية	22776	Latakia	عامية صهيونية	
348662	349882	07/06/2021	Rif Dimashq	فيد الانتظار	1	1000	500	500	0	1	-1		Rif Dimashq	000....		Rif Dimashq	عامية	9069	Idlib	عامية صهيونية		
392615	394273	06/18/2021	Latakia	دفع المخالفة	0	7000	6000	1000	6	10	-4	#####	Latakia	Latakia Tr	689936		Latakia	خاصة	40320	Latakia	عامية 2د	

- Ambiguous Meaning

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S
1	No	Date	Violation	ViolationType	IsPresented	TABN	CarTyt	CarCit	VAL	END_V	POINT	ND_POT	END_DATE	ECity	Billr	License	LicenseI	LicenseT	S
114518	114813	2021-06-15 00:00:00	Damascus	(إلغاء اعتراض	0	415941	عامية	Damascus	5000	2500	4	4	2021-07-31 11:53:32	Damascus					
138285	138654	2021-06-01 00:00:00	Aleppo	(إلغاء اعتراض	0	704800	عامية	Aleppo	7000	3500	6	6	2021-08-01 11:15:44	Aleppo					
206921	207478	2021-06-15 00:00:00	Hama	(إلغاء اعتراض	0	682593	خاصة	Latakia	7000	3500	6	6	2021-09-20 11:38:42	Latakia					
206922	207479	2021-06-15 00:00:00	Hama	(إلغاء اعتراض	0	682593	خاصة	Latakia	7000	3500	6	6	2021-09-20 11:38:49	Latakia					
206923	207480	2021-06-15 00:00:00	Hama	(إلغاء اعتراض	0	682593	خاصة	Latakia	5000	2500	4	4	2021-09-20 11:38:56	Latakia					
400751																			
400752																			

We added a new column to map between null values that we found in primary keys in LicenseNO column and the new column (new primary key). (Every null value takes a specific value).

We needed to loop over 150000 x 85000 elements, takes  $O(n^2)$  and it's an impossible solution so we applied the dictionary data structure property. (Pandas library takes the Data Frame and covert it to a dictionary) in order to use the hashing ability that dictionary provides so we can assign a new primary key to the License Dimension distinct values in  $O(n)$ .

```

newDf["LicenseNO"] = newDf["LicenseNO"].astype("str")
newDf.info()

count = 1
for ind in newDf.index:
    newDf.iloc[ind, 1] = count
    count += 1

sliced_text = slice(0, -2)
for ind in newDf.index:
    newDf.iloc[ind, 0] = newDf.iloc[ind, 0][sliced_text]

dic = newDf.set_index('LicenseNO').to_dict()

dic

```

- Table Modification

Change Data Types:

Pandas dtype	Python type	NumPy type	Usage
object	str or mixed	string_, unicode_, mixed types	Text or mixed numeric and non-numeric values
int64	int	int_, int8, int16, int32, int64, uint8, uint16, uint32, uint64	Integer numbers
float64	float	float_, float16, float32, float64	Floating point numbers
bool	bool	bool_	True/False values
datetime64	NA	datetime64[ns]	Date and time values
timedelta[ns]	NA	NA	Differences between two datetimes
category	NA	NA	Finite list of text values

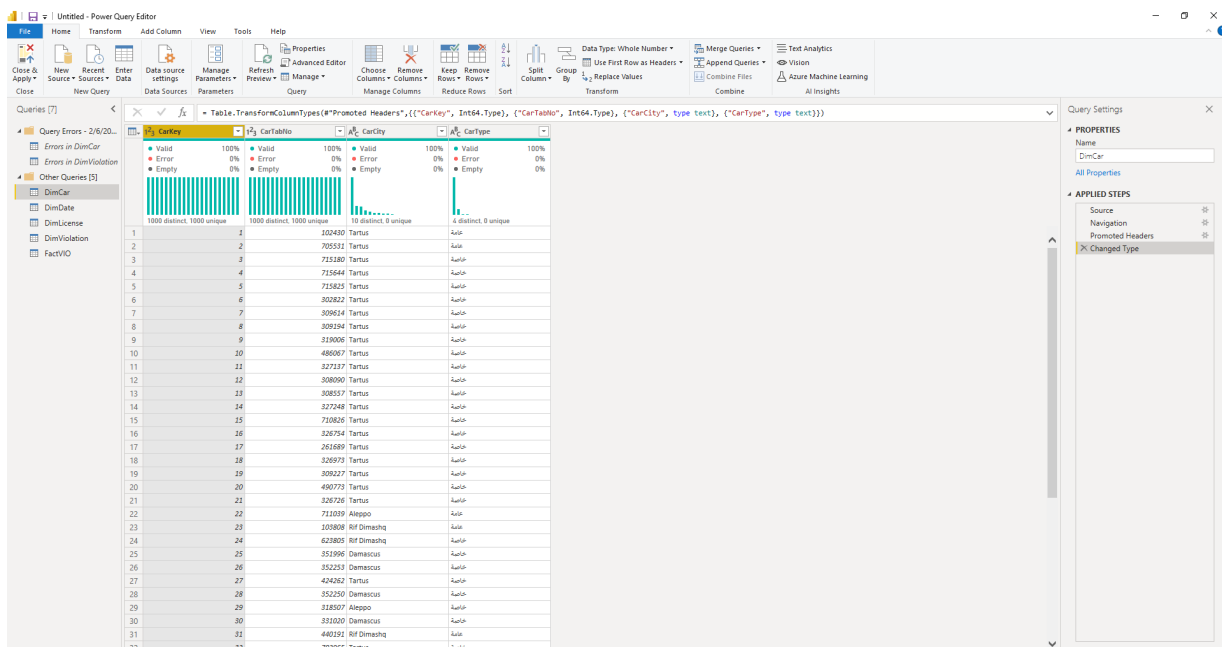


```

1 df["id"] = df["id"].astype("str")
2 df["date_key"] = df["date_key"].astype("str")
3 df["city"] = df["city"].astype("category")
4 df["type"] = df["type"].astype("category")
5 df["is_presented"] = df["is_presented"].astype("bool")
6 df["total_val"] = df["total_val"].astype("int64")
7 df["end_val"] = df["end_val"].astype("int64")
8 df["points"] = df["points"].astype("int64")
9 df["end_points"] = df["end_points"].astype("int64")
10 df["e_city"] = df["e_city"].astype("category")
11 df["biller_name"] = df["biller_name"].astype("category")
12 df["tab_no"] = df["tab_no"].astype("str")
13 df["car_city"] = df["car_city"].astype("category")
14 df["car_type"] = df["car_type"].astype("category")
15 df["lic_id"] = df["lic_id"].astype("str")
16 df["lic_city"] = df["lic_city"].astype("category")
17 df["lic_type"] = df["lic_type"].astype("category")
18
19 df.info()

```

- Data integration while loading to PowrBI

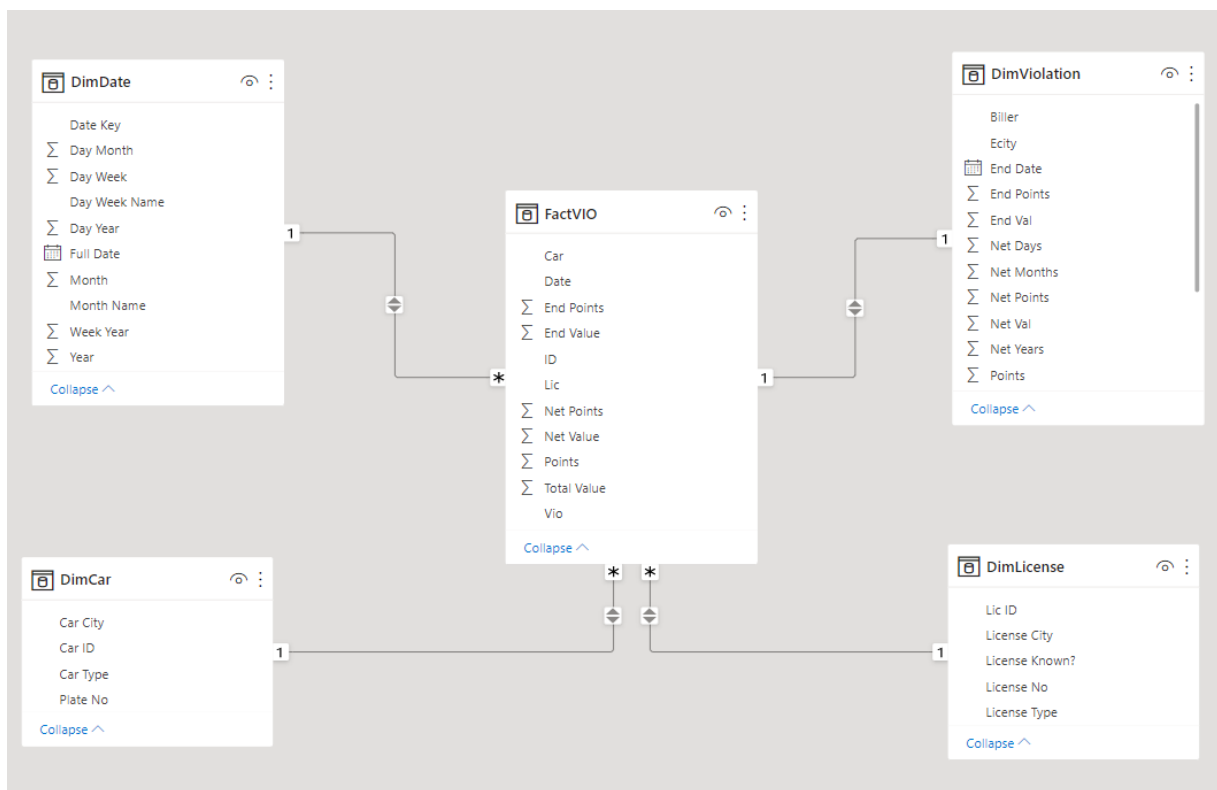


	<div><div></div><div>AB<sub>C</sub> Car ID</div></div>	<div><div></div><div>AB<sub>C</sub> Plate No</div></div>	<div><div></div><div>AB<sub>C</sub> Car City</div></div>	<div><div></div><div>AB<sub>C</sub> Car Type</div></div>
	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>
1	1	102430	Tartus	عامّة
2	2	705531	Tartus	عامّة
3	3	715180	Tartus	خاصّة
4	4	715644	Tartus	خاصّة
5	5	715825	Tartus	خاصّة
6	6	302822	Tartus	خاصّة
7	7	309614	Tartus	خاصّة
8	8	309194	Tartus	خاصّة
9	9	319006	Tartus	خاصّة
10	10	486067	Tartus	خاصّة
11	11	327137	Tartus	خاصّة
12	12	308090	Tartus	خاصّة
13	13	308557	Tartus	خاصّة
14	14	327248	Tartus	خاصّة

	<div><div></div><div>AB<sub>C</sub> License ID</div></div>	<div><div></div><div>AB<sub>C</sub> License No</div></div>	<div><div></div><div>AB<sub>C</sub> License City</div></div>	<div><div></div><div>AB<sub>C</sub> License Type</div></div>	<div><div></div><div>License Is Known</div></div>
	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>	<div> <div>Valid</div> <div>100%</div> </div> <div> <div>Error</div> <div>0%</div> </div> <div> <div>Empty</div> <div>0%</div> </div>
1	1	247717	Hama	خاصّة ب	TRUE
2	2	236555	Tartus	خاصّة ب	TRUE
3	3	165958	Aleppo	خاصّة ب	TRUE
4	4	83358	Tartus	خاصّة ب	TRUE
5	5	208524	Hama	خاصّة ب	TRUE
6	6	147262	Tartus	خاصّة ب	TRUE
7	7	69648	Homs	خاصّة ب	TRUE
8	8	192626	Tartus	خاصّة ب	TRUE
9	9	54403	Latakia	خاصّة ب	TRUE
10	10	64284	Hama	خاصّة ب	TRUE
11	11	238550	Tartus	خاصّة ب	TRUE
12	12	233607	Tartus	خاصّة ب	TRUE
13	13	248671	Tartus	خاصّة ب	TRUE
14	14	5005	Tartus	خاصّة ب	TRUE
15	15	225290	Latakia	خاصّة ب	TRUE
16	16	243241	Tartus	خاصّة ب	TRUE
17	17	229600	Latakia	خاصّة ب	TRUE
18	18	243087	Tartus	خاصّة ب	TRUE
19	19	313623	Aleppo	خاصّة ب	TRUE
20	20	217378	Tartus	خاصّة ب	TRUE
21	21	117598	Latakia	خاصّة ب	TRUE
22	22	250179	Latakia	خاصّة ب	TRUE
23	23	616755	Aleppo	خاصّة ب	TRUE
24	24	213541	Tartus	خاصّة ب	TRUE
25	25	431495	Rif Dimashq	خاصّة ب	TRUE
26	26	382127	Rif Dimashq	خاصّة ب	TRUE

## 2.5. Design the Data Model:

The Data warehouse was built using SQL Server 2019 and we used the star schema to model the DW. Star Schema in data warehouse, in which the center of the star can have one fact table and a number of associated dimension tables. It is known as star schema as its structure resembles a star. The Star Schema data model is the simplest type of Data Warehouse schema. It is also known as Star Join Schema and is optimized for querying large data sets. The design one fact table, which is connected further with multiple dimension tables. The fact table contains all the facts, while the dimension table contains the objects.



### ➤ Why Star Schema?

A star schema is a database organizational structure optimized for use in a data warehouse or business intelligence that uses a single large fact table to store transactional or measured data, and one or more smaller dimensional tables that store attributes about the data. It is called a star schema because the fact table sits at the center of the logical diagram, and the small dimensional

tables branch off to form the points of the star.

A fact table sits at the center of a star schema database, and each star schema database only has a single fact table. The fact table contains the specific measurable (or quantifiable) primary data to be analyzed, such as sales records, logged performance data or financial data. It may be transactional -- in that, rows are added as events happen -- or it may be a snapshot of historical data up to a point in time.

### ➤ **How Star Schema Works?**

The fact table stores two types of information: numeric values and dimension attribute values. Using a sales database as an example:

**Numeric value cells** are unique to each row or data point and do not correlate or relate to data stored in other rows. These might be facts about a transaction, such as an order ID, total amount, net profit, order quantity or exact time.

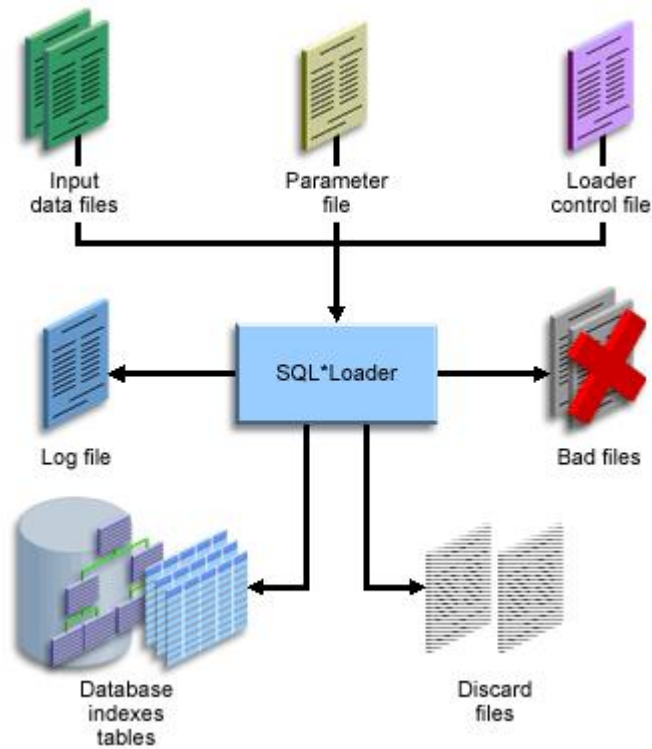
**The dimension attribute values** do not directly store data, but they store the foreign key value for a row in a related dimensional table. Many rows in the fact table will reference this type of information. So, for example, it might store the sales employee ID, a date value, a product ID or a branch office ID.

**Dimension tables** store supporting information to the fact table. Each star schema database has at least one dimension table, but will often have many. Each dimension table will relate to a column in the fact table with a dimension value, and will store additional information about that value.

Constructing a start schema should be carefully done. Each table should have either fact data or dimension data, and avoid mixing the two. Consider the total number of dimension tables to maximize performance. Also, consider the granularity of the data captured to optimize for the types of queries that will be run.

Optimized for querying large data sets, data warehouses and data marts, star schemas support online analytical processing (OLAP) cubes, analytic application, ad hoc queries and business intelligence (BI). They also support count, sum, average and other rapid aggregations

of many fact records. Users can filter and group (sliced and diced) these aggregations by dimensions.



*Figure 2.1 SQL loader environment*

After the data was loaded into the data warehouse, we needed to connect the SQL server DW to the power BI environment to build the dashboards.

## 2.6. Visualization:

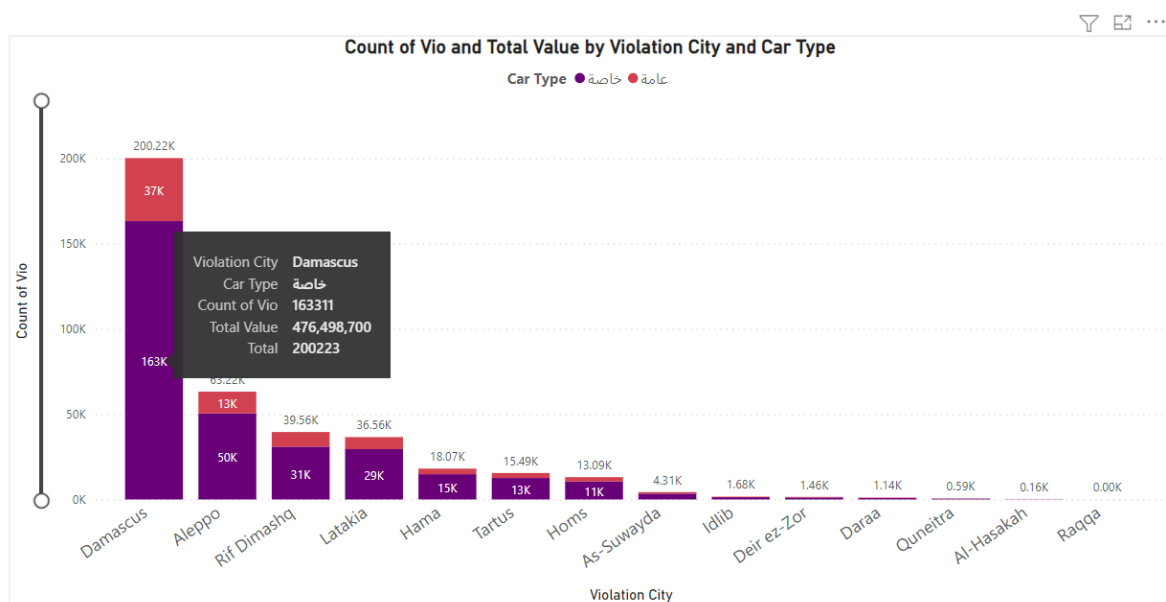
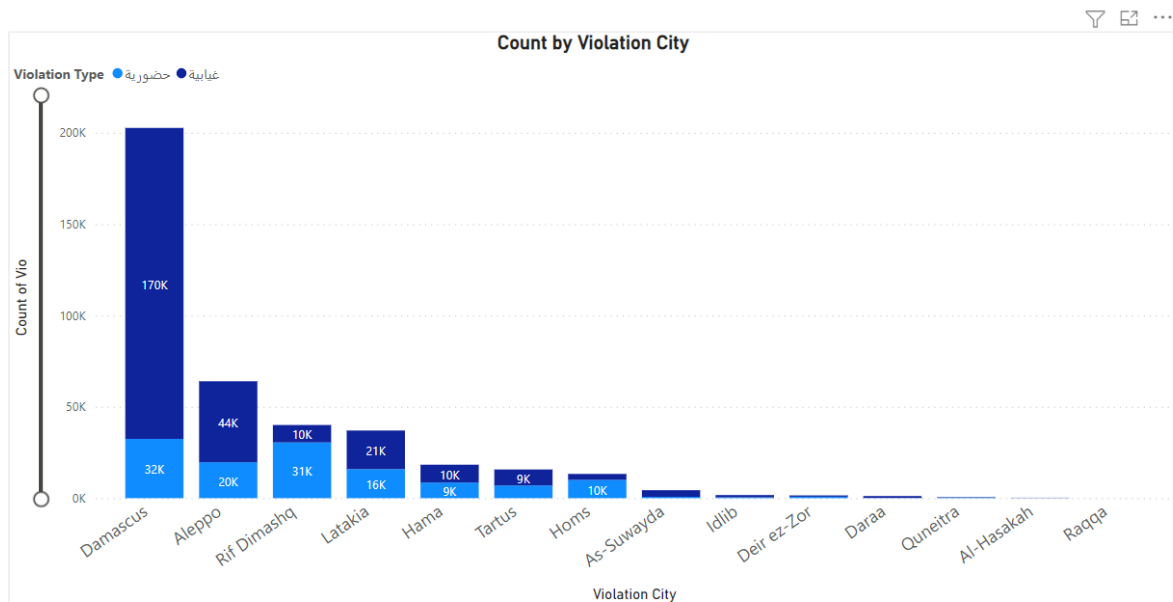
Data Visualization is a process of taking raw data and transforming it into graphical or pictorial representations such as charts, graphs, diagrams, pictures, and videos which explain the data and allow you to gain insights from it. So, users can quickly analyze the data and prepare reports to make business decisions effectively.

### ➤ **What is the importance of Data Visualization?**

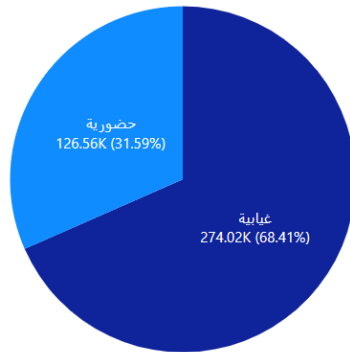
We are inherently in the visual world where pictures or images speak more than words. So, it is easy to visualize a large amount of data using graphs and charts than depending on reports or spreadsheets.

Data visualization is a quick and easy way to convey concepts to the end-users, and you can do experiments with different scenarios by making slight changes.

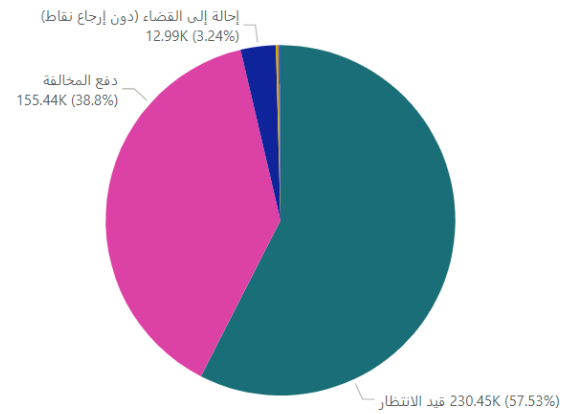
### ➤ **Samples of Our Analysis and Visualizing:**



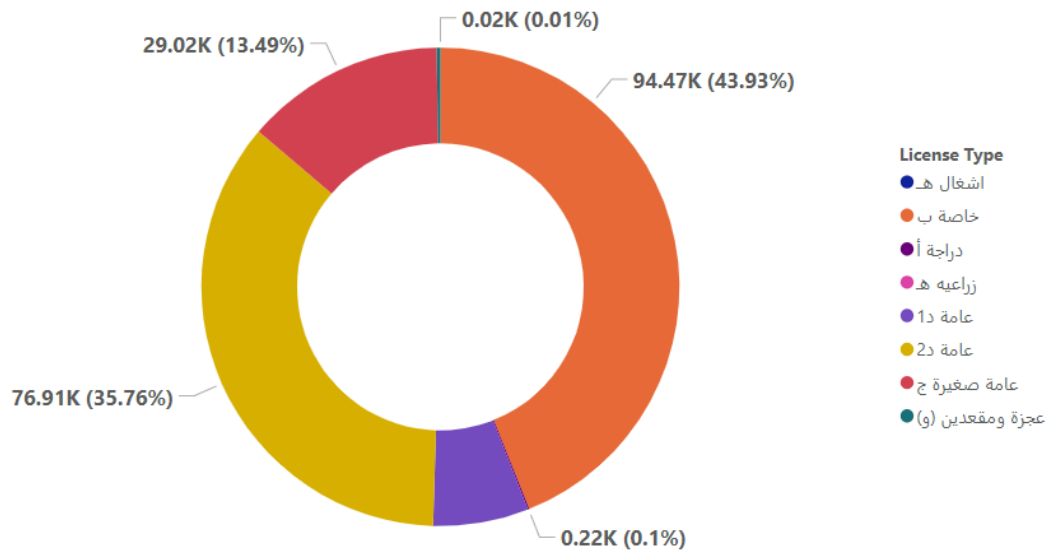
Violation Type



Violation Status

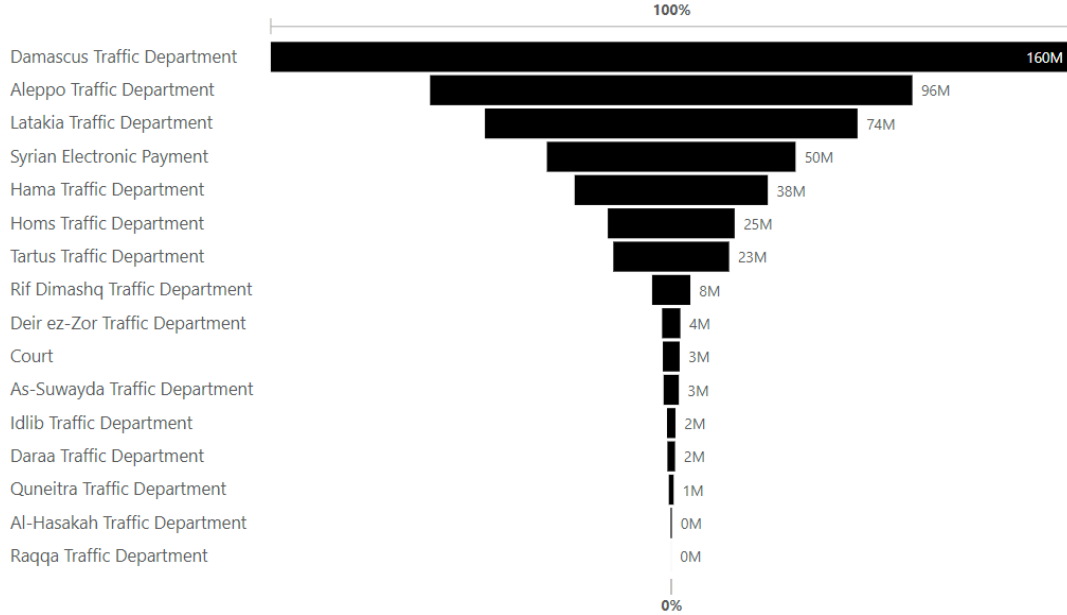


Count by License Type

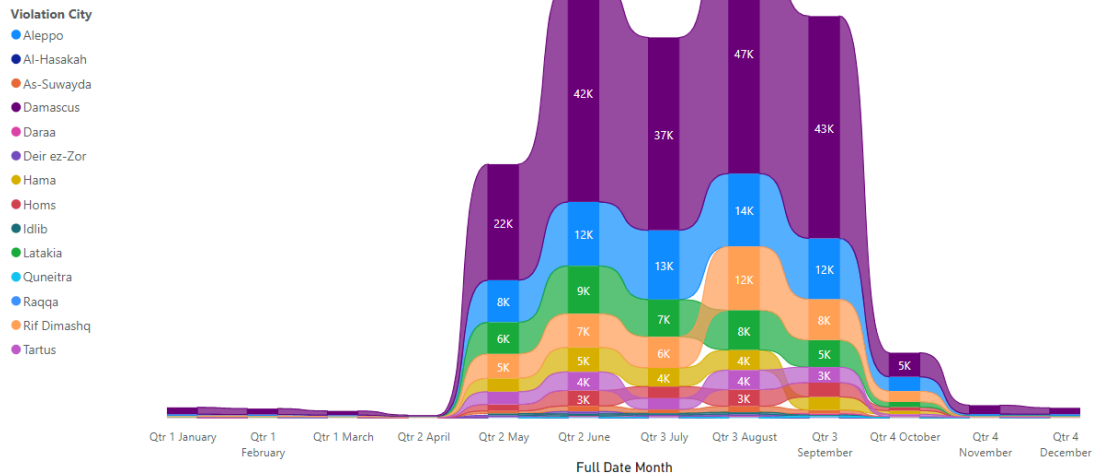


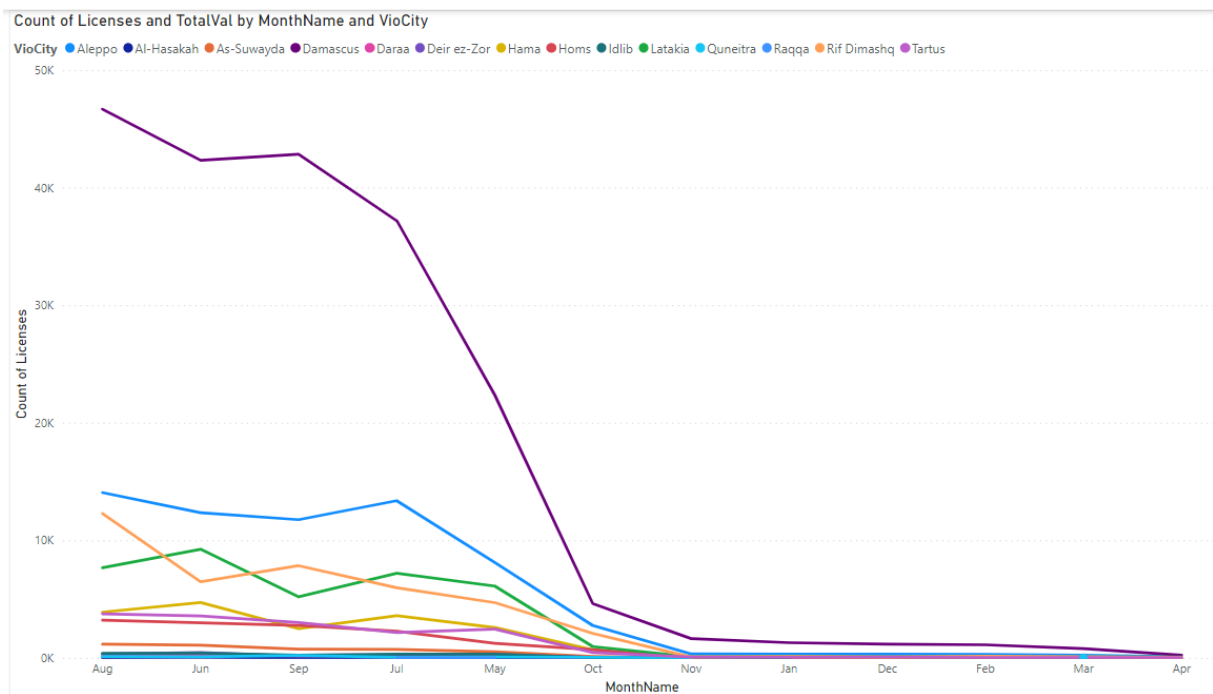
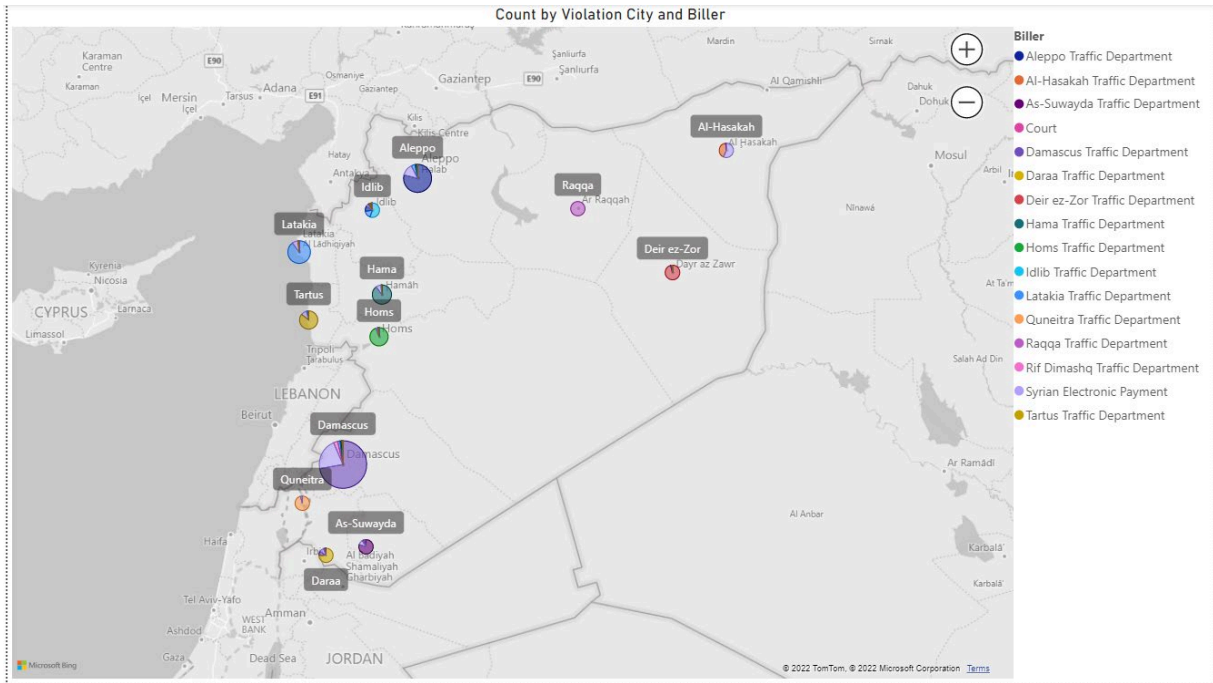


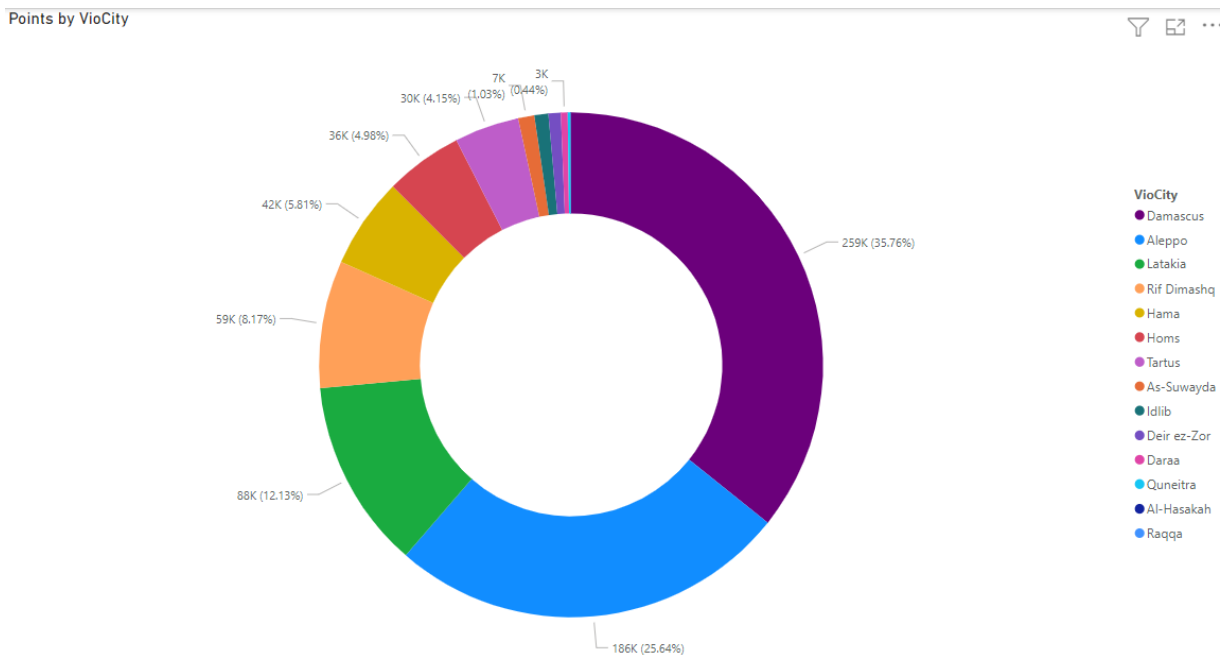
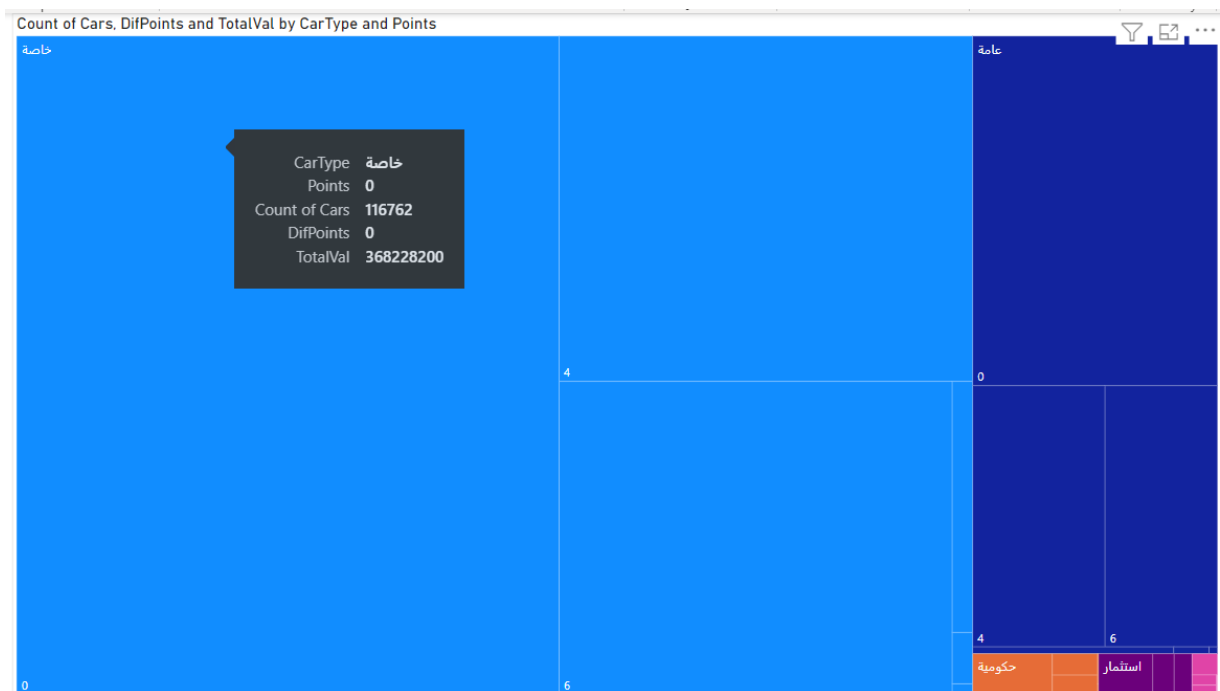
### Total Value by Biller

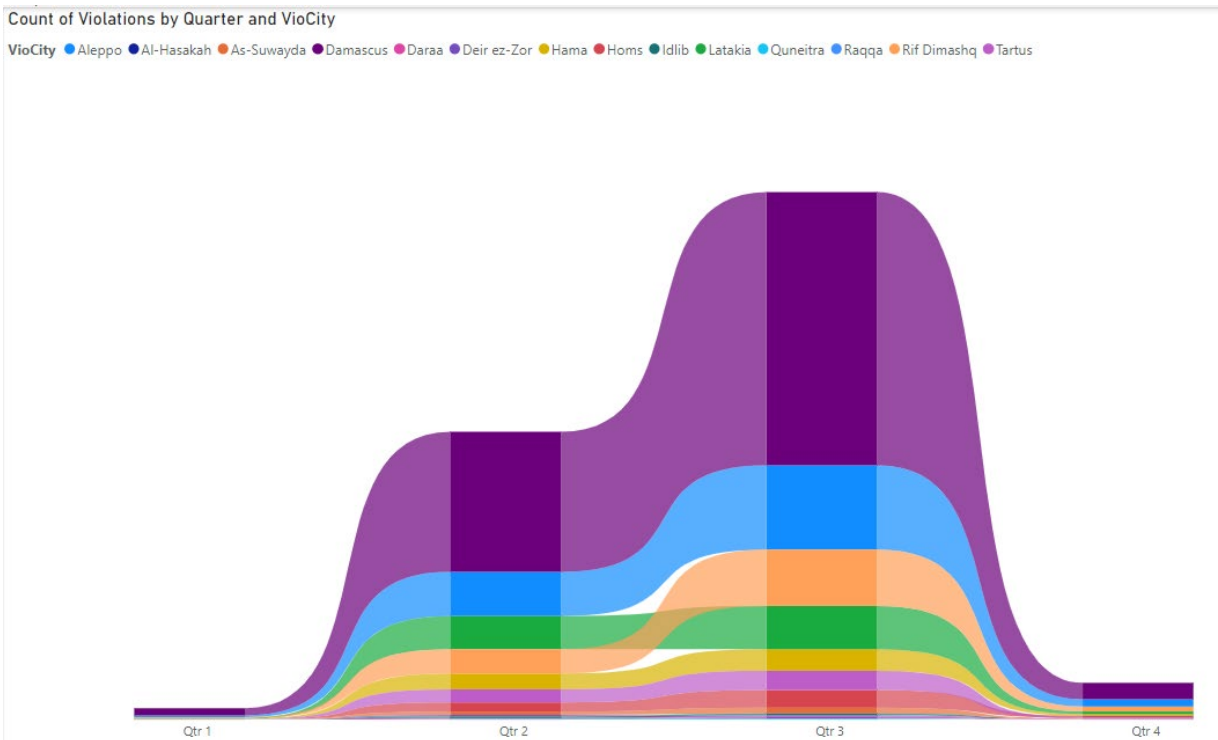
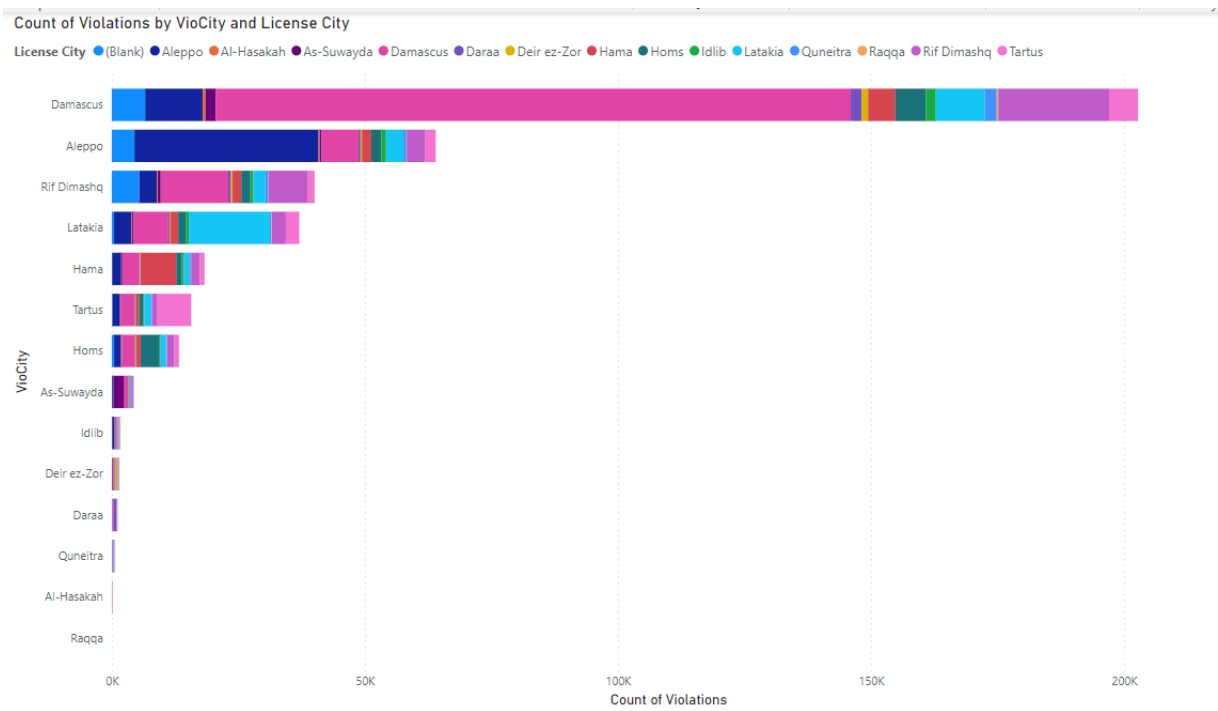


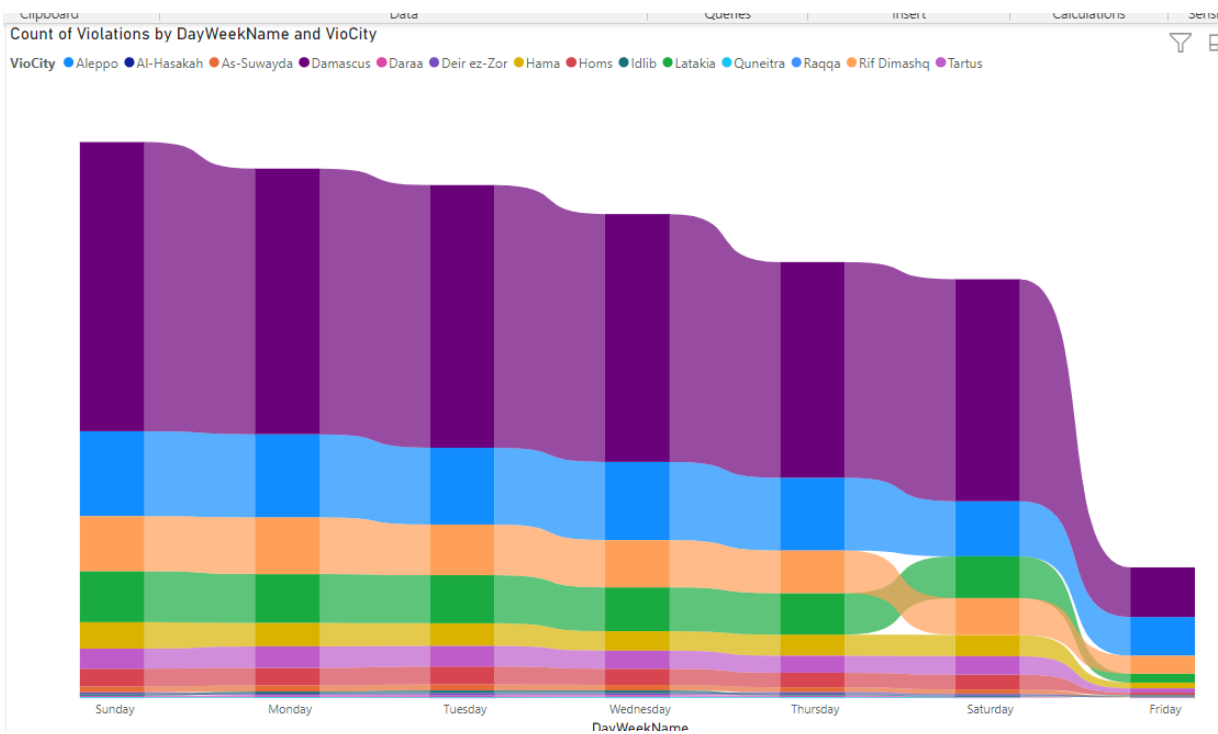
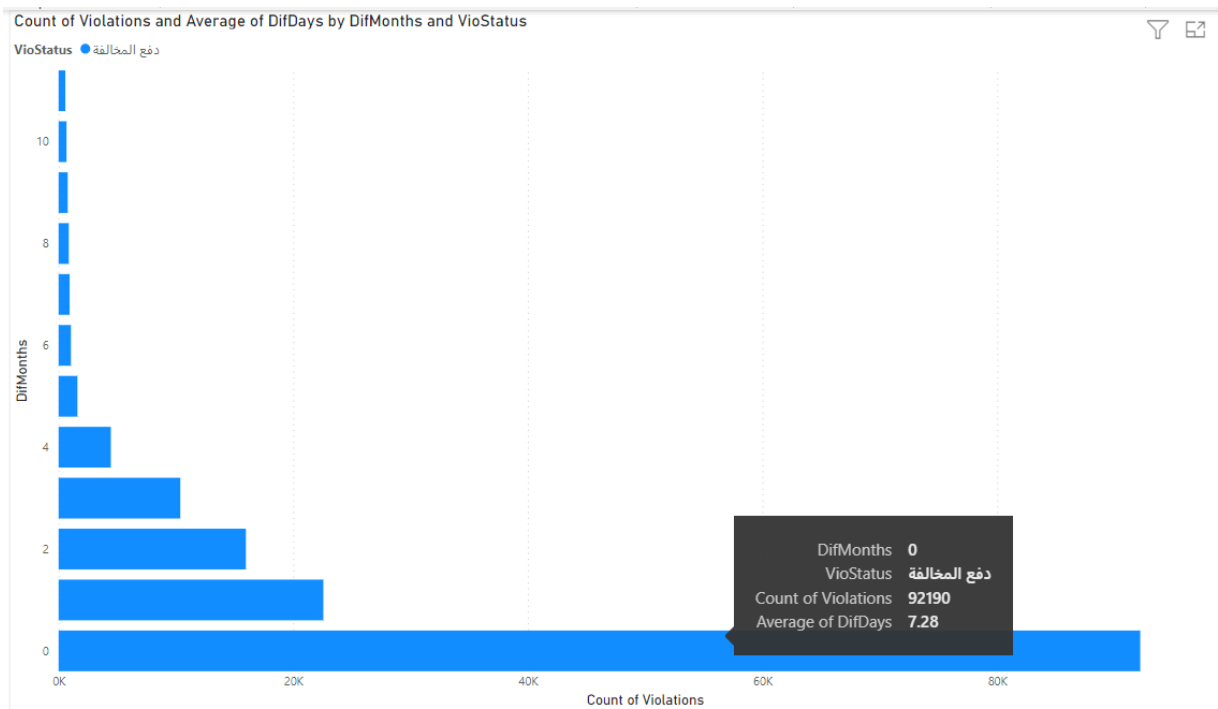
### Count by Quarter and Violation City











## Conclusion:

There are many reasons why companies and ministerial departments adopt BI. Many use it to support functions as diverse as hiring, compliance, production, and marketing. BI is a core business value; it is difficult to find a business area that does not benefit from better information to work with.

Some of the many benefits companies can experience after adopting BI into their business models include faster, more accurate reporting and analysis, improved data quality, better employee satisfaction, reduced costs, and increased revenues, and the ability to make better business decisions. The main objective of this project to show how to apply Business intelligence solutions on the ministry. Performance of the ministry requires continuous monitoring. Therefore, implementing decision support systems and business intelligence technologies, which could provide ministerial officials with summarized data, is considered highly important.

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