# **Alberta A Place to stay: ETL Process**

Algonquin College Course Number: CST\_0022\_10 Database Systems Admin and Mgmt

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### 1. Abstract:

To perform effective analysis on the address data it should be well-structured, and it became easy to perform meaningful analysis using normalized street and city name, leading to more accurate insights for the future use. Normalized street names make precise geospatial analysis like plotting location on maps and calculate the distances and analyze the patterns based on the geographic location in the city. To fulfill these requirements here performed ETL process make huge difference for future analysis.

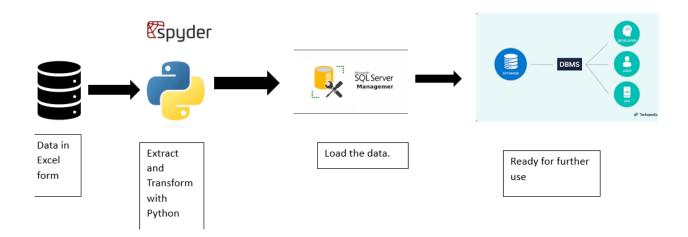
### 2. Introduction:

#### Alberta is the place to stay:

Alberta is the home to six UNESCO designated world Heritage site: the Canadian Rocky Mountain parks, dinosaur Provincial Park, and many more places to explore. [1] It is fourth largest province by area at 661,848 square kilometers, and the fourth most populous place, being home to 4,262,635 people. Having capital as a Edmonton, while Calgary is it's the largest city. The two are Albert's largest census metropolitan areas. [2] Edmonton, the provincial capital of Alberta, is situated roughly in the geographic middle of the province. It is the most northerly metropolitan city in Canada and acts as a crossing point and the hub for the development of resources in the north. The area possesses the majority of West Canada's crude oil processing capacity due to its proximity to the country's main oil reserves. Calgary is surrounded by an extensive grazing region and lies about 280 km (170 mi) south of Edmonton and 240 km (150 mi) north of Montana. The Calgary-Edmonton Corridor is house to about 75% of the province's residents. In the early years of the province, population growth had been encouraged by the land grant policies to the railways. [1] Thus, oil industry and fertile land for the farming of the Albert's state make it distinct stat of the Canada and attractive place to leave within Canada after Ontario. There are several places where humankind is living in huge amount having different areas, street and various types of housing system. Here in this data analysis, we would like to explore home address data of Alberta's residents.

# 3. Tools and techniques:

In this project, first we analyze the data from the excel sheet and explore the data with the help of python to find the dimension, statistical analysis, and the type of the data. From Python script we extract the normalized data from the original data source. Finally, SQL server management studio to load the data in the database with the help of Spyder script. Then load the data which is ready to use for the visualization purpose and further analysis.



### 4. About Data:

This data is extracted from the statistics Canada open database of address (ODA) is a collection of open address point data and available by government of the Canada under the Open Government License-Canada. The inputs for the ODA are datasets provided by municipal, regional, or provincial sources available to the general public through open government portals under various types of open data licenses. The current version of the database (version 1.0) contains approximately 10 million records and includes provinces and territories where open address data were found during the collection period (January to April 2021). Individual datasets sourced from their respective open data portals were processed and harmonized into the ODA. The data sources used do not deploy a uniform standardization system. The ODA harmonizes addresses by converting them to a uniform standard. The bellow's variables included in the ODA are as:

Source Index

**ODA Index** 

**Group Index** 

Civic Number

Street with street type and direction

Street name

Street type

Street direction

Unit

Full address

Postal code

City

**Processed City** 

Standardized street name, type, and direction

Provide

Census subdivision Name and Unique Identifier

Province or Territory Unique Identifier

Latitude and Longitude

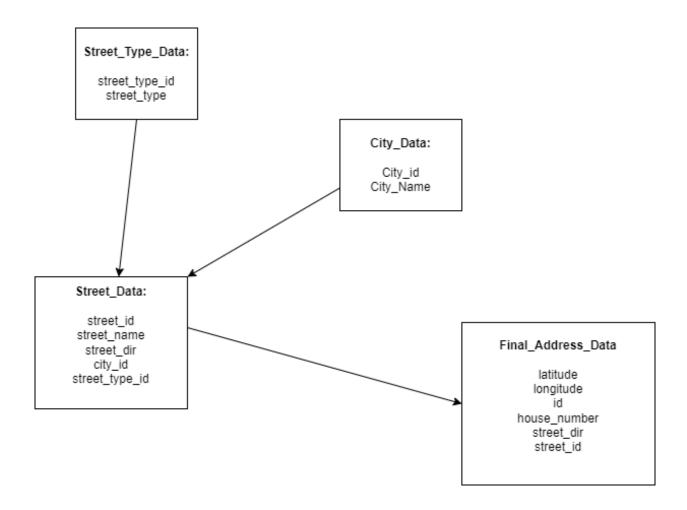
From this data source here, focus on only Alberta province where the data file is in the form of excel and it contain 1777428 rows and 22 columns as below. Thus, data is in 2 dimensions and having size of 39103416.Also, the dataset having all object datatype, where, latitude, longitude and postalcode in float whereas group\_id, csdname and street id are integer format.

The main objective is to find the street id with normalized column with normalized city name according to that street.

## 5. Performed Task:

#### **Extract the Data:**

Here, first we created the relational schema from the original data source to create the needed tables for the further tasks.



As per the below script tables are generated in the SQL server management studio.

We initiated the process by focusing on the unique city entries with the Unique city entries with the dataset. With the help of Unique () function, we extract the distinct city values in the different excel sheet. This function allowed to isolate each city for the further normalizing process.

#### Normalization of the street names:

To enhance the consistency and accuracy of the street name, we employed the normalization process, and it involves transforming the varying street names ensured that data quality and the accuracy of the analysis were upheld.

#### **Transform the Data (Data Preparation):**

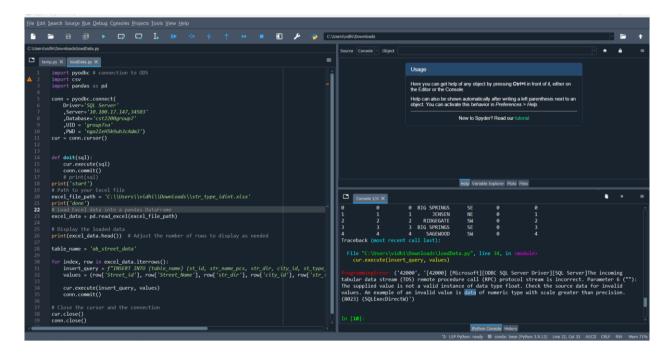
```
▶ for i in street.index:
      fin streetdf.index:
    if(street.str_name_pcs[i] == streetdf.Street_Name[j]):
        street.str_name_id[i] = streetdf.Street_id[j]
        street.str_type_id[i] = streetdf.str_type_id[j]
                 print(i)
  C:\Users\Admin\AppData\Local\Temp\ipykernel_10044\3738504203.py:4: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame
  See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view
  street.str_name_id[i] = streetdf.Street_id[j]
C:\Users\Admin\AppData\Local\Temp\ipykernel_10044\3738504203.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
  See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view
    versus-a-copy
street.str_type_id[i] = streetdf.str_type_id[j]
  45000
  street
                                                                                       str_name_id str_type_id
               latitude longitude street_no str_name_pcs str_type_pcs
  0 51.27301 -113.99282 448 BIG SPRINGS DR
          1 51.29653 -114.00959
                                                15
                                                           JENSEN
                                                                                 CR
          2 51.28530 -114.01526
                                                32 RIDGEGATE
                                                                                 WY
                                                 4 BIG SPRINGS
          4 51.28682 -114.04437
                                             2040 SAGEWOOD
                                                                                 PT
   1777423 50.49860 -111.92423
                                               2 BLUE HERON
                                                                               VIEW
   1777424 50.49830 -111.92430
                                           14 BLUE HERON
                                                                               VIEW
   1777425 50.71807 -111.58501 A 122038 204
                                                                                NaN
   1777426 50.56923 -112.05556 B 155033
   1777427 50.57975 -111.93295 409 145040
  1545126 rows × 7 columns
  streetdf
                                  Street_Name str_dir city_id str_type_id
            0
                        BIG SPRINGS SE 0
                                                                O
                                      JENSEN
                                                     NE
                                                                              1.0
                               RIDGEGATE
                                                                0
                                                                             2.0
                                 BIG SPRINGS
                             SAGEWOOD SW 0
```

Once the city and street name data were successfully extracted but we need to transform the type of the data into one format. To create columns for street and city with one id need to be in one datatype but both have different dimension and the different datatype. For the purpose of normalization, we need unique city id and street type id which was achieved by using python. Thus, we extract the whole data.

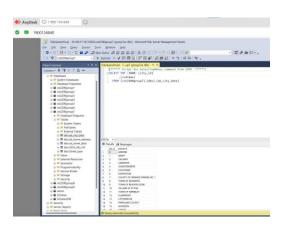
After successfully extracting and normalizing the city and street name data, we combined the data into a comprehensive dataset. Future analysis, reporting, and decision-making will be built upon the data in this collection.

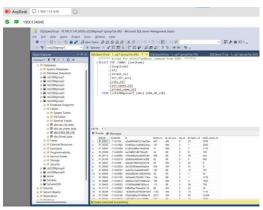
Our method of normalizing street names along with the use of the city\_pcs. unique() function to extract different city values ensures that our dataset is ready for in-depth examination. This initial phase creates the foundation for insightful observations and well-informed choices based on precise and structured facts.

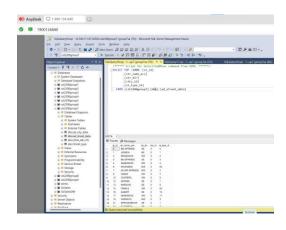
#### **Load the data:**

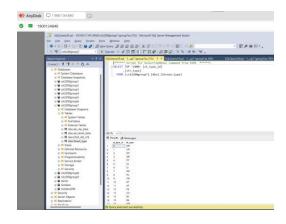


The main SQL file is now ready to do further -process, to clean the data with the help of python. As per the table requirements must need to the normalized data and that would be achieved through python. Now the data is ready to load in the database. Here, Spyder tool is used to load the data into the database.









# 6. Future Work and applications:

After loading the normalized data, one can visualize the data with the help of power BI or Tableau for more visual insight. It can be helpful in city planning, decision making and understanding societal trends. Demographic analysis helps to understand the patterns fro density of the population and the characteristics of different regions. Urban Planning for the government which can identify the high population area and the housing needs for the Alberta region. It can be also useful in the real estate business to understand the housing demand, property values and rental market. Moreover, businesses can use census data to analyze the consumer behavior, tailor market strategies and choose optimal locations for new business based on the population demographics and income level. Thua, overall, after fetching the normalized data it can be useful for the analysis purpose with data visualization to get more insights for the future use.

### 7. Lesson Learned:

After performing these ETL projects we get that if the data is too large as we have in this project then we need more time to load the data. As, here we have 1.5 million rows it took almost more than 28 hours to extract the data from python. So, time management is the key in this kind of project as it is a very time-consuming process. Apart from this, as a new learner, this is not a solo task, group work is very important in such a kind of project where everyone performed their role to complete the project on time and scope as we achieved in this project as a group.

## 8. Conclusion:

In a nutshell, after performing ETL process on conscious data raw data is extracted from the various source and transformed into structured and usable format and finally load by python into the SQL server management studio one can utilized data for the future analysis.

The ETL process's importance is highlighted by its ability to increase data quality, simplify data workflows, and offer enterprises valuable information.

# 9. References:

- 1. <a href="https://en.wikipedia.org/wiki/Alberta#:~:text=Alberta%20is%20home%20to%20six,%2Don%2DStone%20Provincial%20Park">https://en.wikipedia.org/wiki/Alberta#:~:text=Alberta%20is%20home%20to%20six,%2Don%2DStone%20Provincial%20Park</a>.
- "Population and dwelling counts, for Canada, provinces and territories, census metropolitan areas and census agglomerations, 2016 and 2011 censuses – 100% data (Alberta)". Statistics Canada. February 7, 2018. Archived from the original on February 23, 2020. Retrieved December 29, 2020.