LMIA ETL Automation Project Report

1. Project Purpose

The LMIA ETL Automation Project was developed to streamline and automate the process of extracting, transforming, and loading (ETL) Labour Market Impact Assessment (LMIA) data associated with Canada's Temporary Foreign Worker Program (TFWP). The LMIA process is a critical step for Canadian employers who wish to hire foreign workers, as it assesses the impact of such hires on the Canadian labor market. By automating the ETL pipeline, this project aims to provide a reliable, efficient, and repeatable method for preparing LMIA data for analysis and visualization, thereby supporting data-driven decision-making for policymakers, researchers, and employers.

2. Project Workflow and Implementation

2.1 Data Extraction

The project begins by loading raw LMIA data from a CSV file. This step is handled by the load_data function, which reads the CSV into a pandas DataFrame. The data typically contains information about employers, occupations, locations, and the status of LMIA applications.

2.2 Data Transformation

Once loaded, the data undergoes a cleaning process using the clean_data function. This transformation step includes removing duplicate records to ensure each LMIA entry is unique, preventing skewed analysis due to repeated data. Additionally, it handles missing values, specifically by filling missing values in the 'Incorporate Status' column with 'Unknown' to maintain consistency and avoid null-related errors in subsequent analysis.

2.3 Data Loading

After cleaning, the data is saved in two formats:

- Processed CSV file: The cleaned data is exported to a new CSV file for easy sharing and backup.

- SQLite database: The data is also loaded into a SQLite database (Imia_analysis.db) using the save_to_db function. This structured storage allows for efficient querying and integration with analytical tools.

2.4 Data Analysis

The analysis phase retrieves the cleaned data from the SQLite database using the fetch_data function. The basic_report function then generates summary statistics and insights, such as the total number of LMIA applications processed, the distribution of applications by occupation, province, or employer, and trends in LMIA approvals and rejections.

2.5 Data Visualization

The visualization component, implemented in the plot_top_occupations function, creates a bar chart of the top occupations receiving LMIA grants. This visual representation helps users quickly identify which job categories are most in demand for foreign workers in Canada.

3. Methodology

This project follows a modular and automated methodology for data preparation and analysis. The ETL pipeline is implemented using Python, pandas, and SQLite. The extraction step involves loading data from a CSV file. Data transformation is achieved through custom functions that remove duplicates, handle missing values, and ensure data consistency. The cleaned data is stored in both a processed CSV file and a SQLite database for flexible access. Analytical scripts retrieve data from the database to produce summary statistics, while visualization scripts generate informative plots using matplotlib and seaborn. The entire workflow is automated via a master script, allowing for seamless re-execution whenever new data is available.

4. Problem Statement and Solution

4.1 Problem Addressed

Prior to this project, the process of preparing LMIA data for analysis was manual, time-consuming, and error-prone. Data inconsistencies, missing values, and lack of automation made it difficult to generate timely

and accurate reports. This hindered the ability of stakeholders to make informed decisions regarding labor market policies and workforce planning.

4.2 Solution Provided

By automating the ETL process, this project eliminates manual intervention, reduces the risk of human error, and ensures that the data is always up-to-date and analysis-ready. The use of a SQLite database further enhances data accessibility and supports advanced querying for deeper insights.

5. Insights from Visualization

The visualization produced by the project highlights the top occupations that receive the highest number of LMIA grants. This information is crucial for understanding labor shortages and employer needs across Canada. For example, if the visualization shows that certain skilled trades or healthcare positions dominate the LMIA grants, it indicates areas where domestic labor supply may be insufficient, guiding both policy and educational program development.

6. Limitations

While this project delivers a robust and automated approach to LMIA data analysis, several limitations exist. First, the analysis is only as accurate as the data provided; any errors, omissions, or inconsistencies in the source CSV may propagate through the pipeline. Second, the project currently focuses on basic data cleaning and summary statistics. More advanced analyses, such as time series forecasting or geographic visualizations, would require further development. Finally, the SQLite database is suitable for small to medium-sized datasets but may not scale efficiently for very large datasets or concurrent users.

7. Conclusion

The LMIA ETL Automation Project provides a comprehensive, automated solution for preparing, analyzing, and visualizing LMIA data. It supports evidence-based decision-making and helps stakeholders respond proactively to labor market trends in Canada.

8. References

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