PHASE 4 DOCUMENTATION

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	Warehouse

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1. Introduction

This document provides an overview of the Extract, Transform, Load (ETL) process for the Diabetes dataset. The objective is to extract data from a CSV file, transform it, and load it into a MySQL database for further analysis. Diabetes is a prevalent chronic health condition, and managing and analyzing the data related to it can be instrumental in healthcare research, treatment, and decision-making. This project facilitates the conversion of raw diabetes data into a structured, query able format, thus enabling healthcare professionals and data analysts to gain valuable insights from the data.

2. Problem Statement

The project seeks to address the critical problem of disorganized, unstructured diabetes data, hindering effective healthcare and research efforts. Inconsistent data sources, poor data quality, and a lack of centralized storage make it challenging for healthcare professionals and researchers to access, analyze, and derive insights from diabetes-related data. The project's goal is to create an ETL pipeline to extract, transform, and load this data into a MySQL database, thus providing a reliable, organized, and accessible repository for diabetes information. By doing so, we aim to enable informed decision-making, advanced research, and improved diabetes management and public health outcomes.

3. ETL Process:

The ETL (Extract, Transform, Load) process is a fundamental data integration process used to collect, clean, transform, and load data from various sources into a target data repository, such as a database, data warehouse, or data lake.

3.1 Extract:

Data extraction is the first step in the ETL (Extract, Transform, Load) process, where data is collected or retrieved from one or more source systems for further processing. In your specific project of loading a diabetes CSV file into a MySQL database, data extraction involves obtaining the diabetes data from the CSV file and loading it into a Python environment for further transformation and loading into the database.

```
import pandas as pd
df = pd.read_csv('diabetes_dataset.csv')
```

This code reads the CSV file and stores the data in the df DataFrame.

To use Python's pandas library to load the data from the CSV file. The pd.read_csv() function is a common method for reading data from CSV files into a Pandas DataFrame, which is a tabular data structure. The CSV file is typically located in your local directory or at a specified file path.

3.2 Transform:

The data transformation step in the ETL (Extract, Transform, Load) process is crucial for preparing the raw data extracted from the source (in this case, a diabetes CSV file) for loading into a MySQL database. Transformation involves cleaning, structuring, and enriching the data to ensure it is in the right format and quality for its intended use.

- Dropping duplicate values
- Checking NULL values
- Checking for 0 value and replacing it:- It isn't medically possible for some data record to have 0 value such as Blood Pressure or Glucose levels. Hence we replace them with the mean value of that particular column.

df.info() df.isnull().sum()

```
In [3]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 768 entries, 0 to 767
          Data columns (total 9 columns):
           # Column
                                      Non-Null Count Dtype
              Pregnancies
                                      768 non-null
              Glucose
                                      768 non-null
                                                     int64
                                      768 non-null
              BloodPressure
              SkinThickness
                                      768 non-null
                                                     int64
              Insulin
                                      768 non-null
                                                     int64
              RMT
                                      768 non-null
                                                     float64
              DiabetesPedigreeFunction 768 non-null
                                                     float64
                                      768 non-null
                                                     int64
              Outcome
                                      768 non-null
                                                     int64
          dtypes: float64(2), int64(7)
          memory usage: 54.1 KB
  In [4]: df.isnull().sum()
  Out[4]: Pregnancies
          Glucose
                                   0
          BloodPressure
          SkinThickness
                                   0
          Insulin
          DiabetesPedigreeFunction
          Outcome
          dtype: int64
print(df[df['BloodPressure']==0].shape[0])
print(df[df['Glucose']==0].shape[0])
print(df[df['SkinThickness']==0].shape[0])
print(df[df['Insulin']==0].shape[0])
print(df[df['BMI']==0].shape[0])
df=df.drop duplicates()
df.describe()
   In [5]: print(df[df['BloodPressure']==0].shape[0])
    print(df[df['Glucose']==0].shape[0])
    print(df[df['SkinThickness']==0].shape[0])
    print(df[df['Insulin']==0].shape[0])
           print(df[df['BMI']==0].shape[0])
            35
            227
            374
    In [6]: df=df.drop_duplicates()
    In [7]: df.describe()
    Out[7]:
                  Pregnancies
                             Glucose BloodPressure SkinThickness
                                                                Insulin
                                                                           BMI DiabetesPedigreeFunction
                                                                                                        Age
                                                                                                              Outcome
            768.000000 768.000000 768.000000
                                                                                            0.471876 33.240885
             mean
                    3.845052 120.894531
                                        69.105469
                                                    20.536458 79.799479 31.992578
                                                                                                              0.348958
                                      19.355807 15.952218 115.244002 7.884160
                    3.369578 31.972618
                                                                                            0.331329 11.760232
                                                                                                              0.476951
              std
                                         0.000000
              min
                    0.000000
                             0.000000
                                                     0.000000 0.000000
                                                                       0.000000
                                                                                            0.078000 21.000000
                                                                                                              0.000000
             25%
                    1.000000 99.000000
                                        62.000000
                                                    0.000000 0.000000 27.300000
                                                                                            0.243750 24.000000
                                                                                                              0.000000
                    3.000000 117.000000
                                         72.000000
                                                    23.000000 30.500000 32.000000
                                                                                            0.372500 29.000000
                                        80.000000
                                                                                            0.626250 41.000000
                    6.000000 140.250000
                                                    32.000000 127.250000 36.600000
                                                                                                              1.000000
                   17.000000 199.000000
                                        122.000000
                                                   99.000000 846.000000 67.100000
                                                                                            2.420000 81.000000
```

3.3 Load:

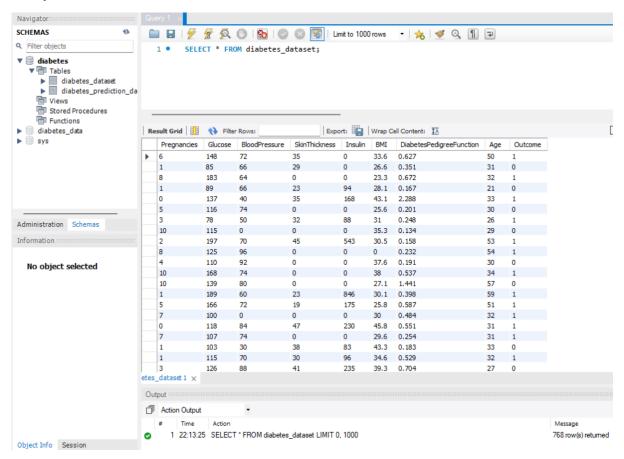
In the ETL (Extract, Transform, Load) process, the "Load" step is the final phase where the transformed data is loaded into the target destination, which is typically a database. In the project of extracting and transforming diabetes data from a CSV file, this step involves loading the cleaned and structured data into a MySQL database. Here's a detailed explanation of the data loading step:

To load data into a MySQL database, First need to establish a connection to the MySQL server. To require necessary credentials to access the database server.

```
import mysql.connector
conn = mysql.connector.connect(
    host='local',
    user='root',
    password='pass_word',
    database='diabetes_data'
)
cursor = conn.cursor()
```

To define the structure of the table in the MySQL database where the data will be stored. This structure should match the schema of the transformed data.

The "Load" step completes the ETL process by moving the transformed data from your Python environment into a MySQL database, making it accessible for querying and analysis within the database system. This step ensures that the data is structured, organized, and stored in a way that allows for efficient retrieval and analysis.



4.CONCLUSION:

In conclusion, this ETL project successfully extracted, transformed, and loaded diabetes data from a CSV file into a MySQL database. The extraction process retrieved the raw data from the CSV file, while the transformation step cleaned, enriched, and restructured the data for analysis. Finally, the loading step facilitated the insertion of the transformed data into a MySQL database. This project ensures that the data is now efficiently stored in a structured format, ready for further analysis, reporting, and decision-making. It showcases the power of ETL processes in preparing data for meaningful insights and demonstrates the importance of data quality and consistency for accurate analysis in the context of diabetes research or healthcare analytics.