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# Pulse to PostgreSQL - Code Explanation

This document provides a detailed explanation of the <a href="mailto:pulse\_to\_postgresql.py">pulse\_to\_postgresql.py</a> script, which extracts data from a Microsoft SQL Server database and loads it into a PostgreSQL database.

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# **Imports**

```
import pandas as pd
from datetime import datetime, timedelta
from sqlalchemy import create_engine, text
import pyodbc
import psycopg2
import sys
import os
from sqlalchemy.exc import SQLAlchemyError
```

- pandas (pd): Used for data manipulation and analysis
- datetime, timedelta: For handling date and time operations
- sqlalchemy: SQL toolkit and ORM library used to communicate with databases
- pyodbc: Library for connecting to ODBC databases (MS SQL Server in this case)
- psycopg2: PostgreSQL database adapter
- sys, os: Standard Python modules for system interactions
- **SQLAIchemyError**: For handling sqlalchemy-specific exceptions

# **PulseDBTransformer Class**

## Constructor

The constructor initializes the following:

```
def __init__(self, pg_host='localhost', pg_port=5432, pg_dbname='em_pulse_data',
pg_user='postgres', pg_password='postgres'):
```

#### **SQL Server Connection**

```
# SQL Server connection parameters
self.server = '10.20.2.10'
self.database = 'pulse'
self.username = 'Pulse_RO'
self.password = 'PD@T@r3@der'
self.connection_string = f"DRIVER={{ODBC Driver 17 for SQL Server}};SERVER=
{self.server};DATABASE={self.database};UID={self.username};PWD={self.password}"
```

```
self.engine = create_engine("mssql+pyodbc:///?odbc_connect=" +
self.connection_string)
```

This section sets up the connection to the source SQL Server database:

- Defines the server address, database name, username, and password
- · Creates a connection string compatible with the ODBC driver
- Initializes a SQLAlchemy engine for database interactions

### **PostgreSQL Connection**

```
# Target PostgreSQL connection parameters
self.pg_host = pg_host
self.pg_port = pg_port
self.pg_dbname = pg_dbname
self.pg_user = pg_user
self.pg_password = pg_password
self.pg_engine = create_engine(f"postgresql://{pg_user}:{pg_password}@{pg_host}:
{pg_port}/{pg_dbname}")
```

This section sets up the connection to the target PostgreSQL database:

- Takes connection parameters as constructor arguments with default values
- Creates a SQLAlchemy engine for PostgreSQL database interactions

## Mill Names and Configuration

This defines the list of mills to process.

## **SQL Tags Dictionary**

```
# SQL tags dictionary mapping tag IDs to mill names for each feature
self.sql_tags = {
   'Ore': {"485" : "Mill01", "488" : "Mill02", ...},
   'WaterMill': {"560" : "Mill01", "580" : "Mill02", ...},
```

```
····
}
```

This dictionary maps numeric tag IDs in the SQL Server to the corresponding mill and feature. The structure is:

- Top level: Feature names (Ore, WaterMill, etc.)
- Second level: Tag ID (as string) → Mill name

#### **Table Names**

This defines which tables to query in the SQL Server database:

- · Default is just the current data in 'LoggerValues'
- · Commented section allows processing historical archived data

### **Filter Timestamp**

```
# Initialize timestamp filter to None (get all data)
self.filter_timestamp = None
```

This parameter is used for incremental loading in the append operation:

- When None, all available data is processed
- · When set to a timestamp, only data newer than that timestamp is processed

# **Data Processing Methods**

## read\_sql\_table

```
def read_sql_table(self, table_name, feature):
```

```
"""Read data from SQL Server for a specific feature"""
```

#### This method:

- 1. Builds a SQL query for a specific feature from a specific table
- 2. Adds timestamp filtering if applicable
- 3. Executes the query and retrieves the data
- 4. Processes the data:
  - Removes duplicate timestamps
  - Pivots the data (IndexTime as index, LoggerTagID as columns)
  - Fills missing values
  - Resamples to 1-minute intervals
  - Renames columns to mill names based on the sql\_tags dictionary
- 5. Returns a DataFrame with mills as columns and timestamps as index

### compose\_feature

```
def compose_feature(self, feature):
    """Combine data from all tables for a specific feature"""
```

#### This method:

- 1. Processes each table in self.table\_names for the given feature
- 2. Combines the results into a single DataFrame
- 3. Sorts by timestamp and removes duplicate timestamps
- 4. Shifts the timestamps by 2 hours (to align data)
- 5. Fills missing values
- 6. Returns a combined DataFrame for the feature

## create\_mill\_dataframe

```
def create_mill_dataframe(self, mill):
    """Create a dataframe for a specific mill with all features"""
```

#### This method:

- 1. Processes each feature for the specified mill
- 2. Finds the common time index across all features

- 3. Creates a DataFrame with features as columns for the mill
- 4. Returns a DataFrame with features as columns and timestamps as index

# **Database Operations**

## save\_to\_postgresql

```
def save_to_postgresql(self, schema='mills'):
    """Save all mill data to PostgreSQL database, replacing existing data"""
```

#### This method:

- 1. Creates the specified schema if it doesn't exist
- 2. Processes each mill in self.mills
- Creates a DataFrame for each mill with all features
- 4. Formats the table name (e.g., Mill01 → MILL 01)
- 5. Saves each mill's data to PostgreSQL, replacing any existing data
- 6. Reports progress and results

### append\_to\_postgresql

```
def append_to_postgresql(self, schema='mills'):
    """Append new data to existing PostgreSQL database tables, only adding records
newer than the latest timestamp"""
```

### This method implements incremental loading:

- 1. Creates the specified schema if it doesn't exist
- 2. Stores the original table\_names to restore later
- 3. For each mill:
  - Checks if the table exists in PostgreSQL
  - If it exists, finds the latest timestamp in the table
  - Sets a filter timestamp to only process newer data (with a 2.1-hour buffer for time shifts)
  - Creates a DataFrame for the mill with the filtered data
  - Filters out data with timestamps not newer than the latest timestamp
  - Appends the new data to the existing table

- If the table doesn't exist, creates it with all available data
- 4. Restores the original state (clears filter timestamp)
- 5. Reports progress and results

## **Main Function**

```
def main():
   # PostgreSQL connection parameters
    pg_host = 'em-m-db4.ellatzite-med.com'
    pg_port = 5432
    pg_dbname = 'em_pulse_data'
    pg_user = 's.lyubenov'
    pg_password = 'tP9uB7sH7mK6zA7t'
    # Initialize the transformer with PostgreSQL connection parameters
    transformer = PulseDBTransformer(
       pg_host=pg_host,
       pg_port=pg_port,
       pg_dbname=pg_dbname,
       pg_user=pg_user,
       pg_password=pg_password
    )
    # Save or append based on command line argument
    if len(sys.argv) > 1 and sys.argv[1] == 'append':
        transformer.append_to_postgresql()
    else:
        transformer.save_to_postgresql()
```

#### The main function:

- 1. Defines PostgreSQL connection parameters
- 2. Initializes the PulseDBTransformer class with these parameters
- 3. Determines whether to perform a full save or incremental append operation:
  - If the script is run with the 'append' argument, it performs an append operation
  - o Otherwise, it performs a full save/replace operation

## **Data Flow**

1. Extract: Data is extracted from SQL Server by:

- Querying tables listed in self.table\_names
- Filtering by tag IDs in the self.sql tags dictionary
- Optional timestamp filtering for incremental loading

## 2. Transform: Data is transformed by:

- Pivoting to organize by timestamp and tag ID
- Renaming columns to meaningful mill names
- Resampling for consistent time intervals
- Filling missing values
- Time shifting (2 hours)
- Organizing into mill-specific DataFrames with features as columns

## 3. Load: Data is loaded into PostgreSQL by:

- Creating schemas and tables as needed
- Either replacing existing data or appending new data
- Using SQLAlchemy's to sql method for the actual loading