ETL Technical Manual

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UW ETL Processing Final

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1. Introduction

1.1 Purpose of the Document

This document serves as a comprehensive technical guide for understanding and implementing ETL (Extract, Transform, Load) processes, of a small medical clinic data analytics project. The project involves automating the data ingestion, transformation, and loading processes from CSV files to a structured Data Warehouse. The ETL solution prioritizes scalability, efficiency, and maintainability using SQL Server, SSIS, and Python. This manual provides step-by-step instructions on ETL development, best practices, and troubleshooting techniques.

Additionally, it outlines the four milestones of the project, covering file imports, data warehouse integration, non-SQL ETL processes, and automation. Security, compliance, and performance optimization strategies are also discussed to ensure smooth data flow between systems. Whether building an ETL pipeline for the first time or optimizing an existing one, this document serves as a valuable reference for designing scalable and efficient ETL solutions.

Current Situation

Currently, individual clinics send daily CSV files to the corporate office, where they are manually reviewed, cleaned, and added to one of two databases. This manual ETL process is time-consuming, prone to human error, and lacks scalability, especially as the number of clinics increases. The process delays data availability, affecting decision-making and efficiency.

2. Project Overview

The ETL project automates data movement from daily CSV uploads into structured databases, providing analytical insights. The existing system is entirely manual, and this solution introduces automation to streamline data integration.

- Source Systems: CSV files from clinics
- Destination Systems: Patients and DoctorsSchedules databases,
 DWClinicReportData (Data Warehouse), and Excel reports
- **ETL Tools**: SQL Server, SSIS, Python
- Deployment: Visual Studio-based structured solution

2.1 Current Workflow over view

Currently, the business has individual clinics send data to a corporate office by uploading CSV files each day. Those **files are then added to one of the two databases**. The <u>current</u> ETL process is **entirely manual**.

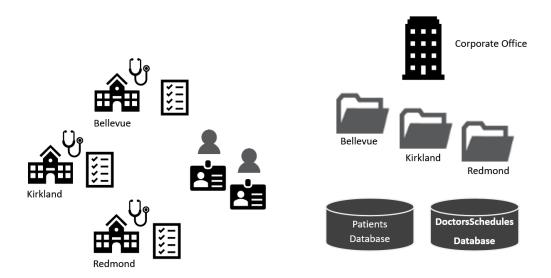


Figure 1. Topology of the current design.

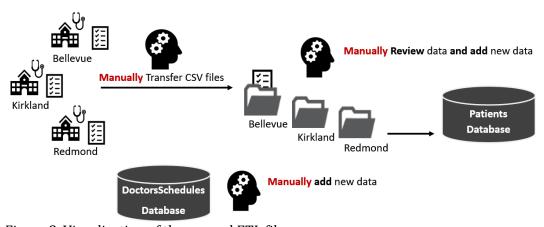


Figure 2. Visualization of the manual ETL file process

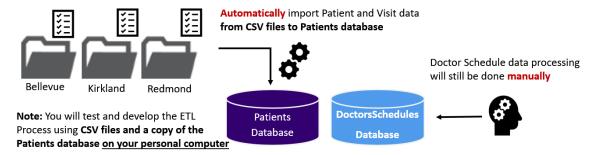


Figure 3. Visualization of the manual ETL file process

2.2, Proposed ETL process

A new ETL process will automate parts of this workflow, with potential for full automation in the future. Below, you'll find detailed descriptions of the entire process and steps, along with sample code snippets and screenshots for clarity.

2.3 Audience

This manual is intended for a wide range of professionals involved in data management and analytics, including:

- ETL Developers Those responsible for designing and implementing ETL pipelines.
- Data Engineers Professionals who manage data pipelines and ensure data flows correctly.
- Database Administrators Responsible for maintaining data integrity and optimizing database performance.
- Business Intelligence Analysts Those who analyze and report on business data.
- IT Managers Decision-makers overseeing data-related projects.

This document assumes some familiarity with databases, SQL, and basic scripting but does not require advanced ETL experience.

2.4 ETL Overview

2.4.1 What is ETL?

ETL stands for Extract, Transform, and Load—a three-step process used for integrating data from multiple sources into a centralized data warehouse or analytics system. The Extract

phase retrieves data from various sources such as databases, web APIs, and flat files. The Transform phase applies business rules to clean, standardize, and enrich the data. The Load phase ensures the data is stored in the target system, ready for reporting and analysis. ETL is widely used in data warehousing, analytics, and business intelligence, ensuring that organizations work with high-quality, consistent, and structured data.

2.4.1 Why ETL is Important?

ETL plays a crucial role in modern data-driven organizations by enabling them to consolidate data from multiple sources and prepare it for analysis. Without ETL, businesses would struggle with inconsistent, incomplete, and redundant data. ETL ensures that data is cleaned, standardized, and formatted properly before it reaches reporting tools. Additionally, ETL processes support historical data storage, allowing companies to perform trend analysis, forecasting, and strategic decision-making. ETL also enhances data governance and compliance, ensuring that sensitive information is processed securely according to industry regulations such as GDPR and HIPAA.

3.Environment Setup

Before beginning ETL development, ensure the following components are installed and configured:

- SQL Server with Management Studio
- Visual Studio with SSIS extensions
- Python 3.9 with required libraries
- Access to database backup files (.BAK) for restoration
- CSV files for testing in the C:_BISolutions\ClinicDailyData directory

4. Milestone 1: File-Based ETL

Data Review

 Assess CSV file structures for inconsistencies (e.g., column names, data types, missing values).

Setting Up the Development Environment

• Create a Visual Studio solution with organized folders.

Database Restoration

Execute the following SQL script to restore databases:

ALTER DATABASE [Patients] SET SINGLE_USER WITH ROLLBACK IMMEDIATE;

RESTORE DATABASE [Patients]

FROM DISK = N'C:/_BISolutions/Databases/Patients.bak'

WITH RECOVERY, REPLACE;

ALTER DATABASE [Patients] SET MULTI_USER;

GO

Data Transformation Documentation

Maintain an Excel metadata sheet tracking transformations.

SSIS Package Implementation

• Develop an SSIS package (ETLFilesToDatabases.dtsx) to automate CSV imports.

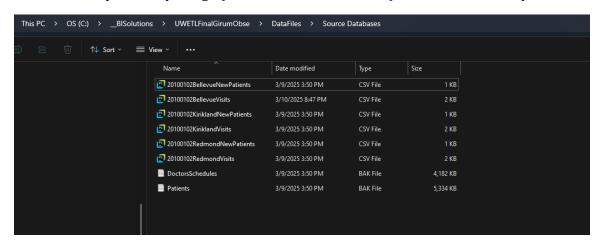


Figure 4 Excel source file for different clinics

Figure 5, sql code script showing restoring the patients and doctorschedule database

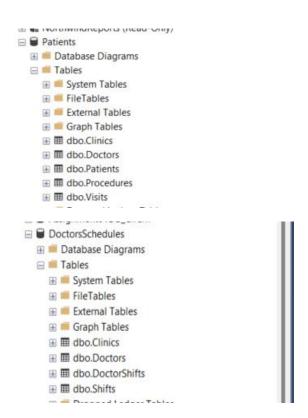


Figure 6, showing doctorschedule and patients database tables

Figure 7: part of sql script showing how data loaded to databases

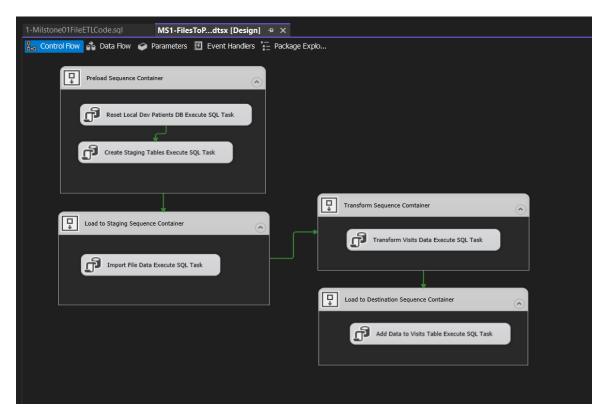


Figure 8, ssis package for dataloaded from excel source to visits table in patients database

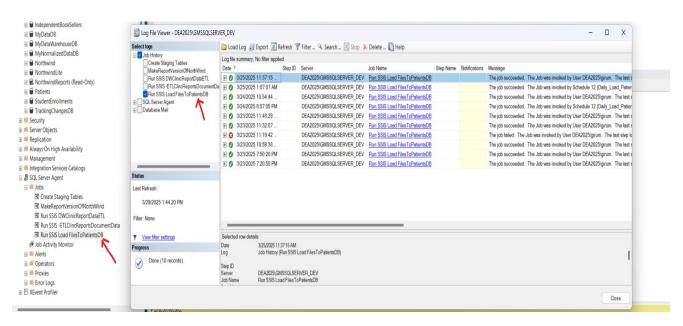


Figure 9, showing screenshot of scheduled job to load data to patients database tables

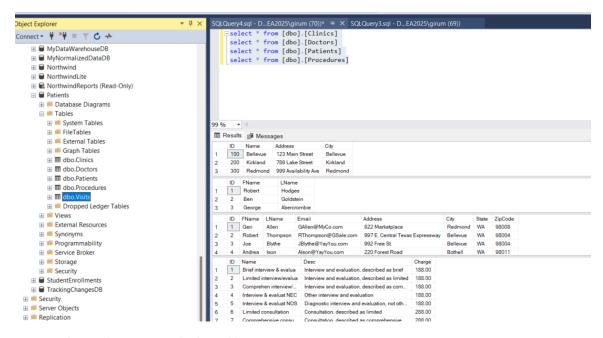


Figure 10, showing data in patients database tables

5. Milestone 2: Data Warehouse ETL

Process Steps

Creating the Data Warehouse

- Run the provided SQL script (`\Scripts\CreateDWClinicReportData.sql`) to create the **DWClinicReportData** database.

```
ETL Final Preject: DMCLinicReportData_GirumO

Dev: MROCCA
Dev: PROCCA
ChangeLog: Other, When, What)

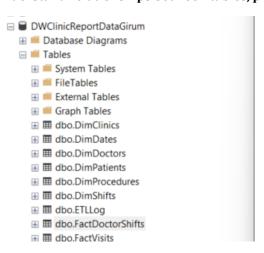
Renot, 37/1/17, removed addresses from DisPatients
Renot, 37/1/17, addred SCD columns to DisPatients
Renot, 3
```

Figure 11, sql code script of the datawarehouse

- Verify that all required tables, schemas, and constraints are properly created.
- Check for any errors or missing objects in the database after execution.

Reviewing the Database Design

- Identify the **dimension tables** (e.g., DimPatients, DimDoctors, DimShifts) and fact tables(e.g., FactDoctorShifts and FactVisits).
- Understand relationships between tables, primary and foreign keys.



SQL ETL and SSIS Implementation

- **Create SQL Views and Stored Procedures**:
- Develop **views** to extract necessary data from the **DoctorsSchedules** and **Patients** OLTP databases.
- Write **stored procedures** to transform and load the data into the **DWClinicReportData** warehouse.
- Implement **data cleansing** to remove duplicates, handle null values, and ensure consistency.

Figure 12, showing ETL script of data load from OLTP database to the OLAP database example of Dimdoctor table

- **Develop an SSIS Package (DWClinicReportDataETL.dtsx)**:
- Create an **SSIS package** to automate the ETL process.
- Configure **Data Flow tasks** to extract data from the source OLTP databases.
- Use **Staging Tables** for incremental loads and ensure proper error handling.
- Implement **Logging and Notifications** to track process execution and failures.

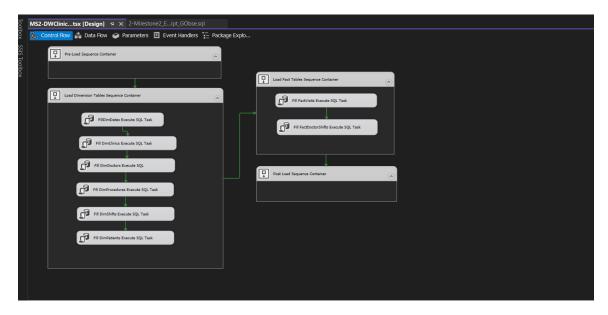


Figure 13, showing ssis package of the datawarehouse

- Schedule the SSIS package for execution via **SQL Server Agent** or manual triggers.

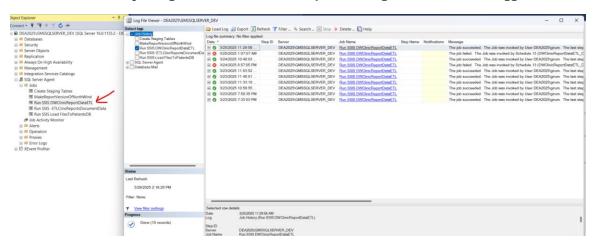


Figure 14, showing the ssis scheduled job and job history.

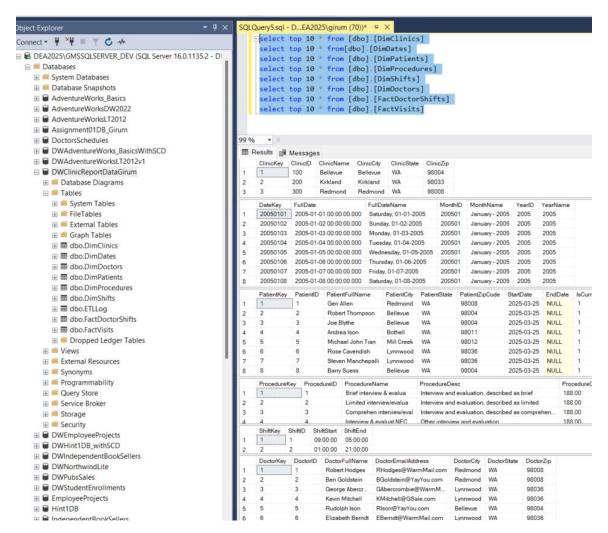
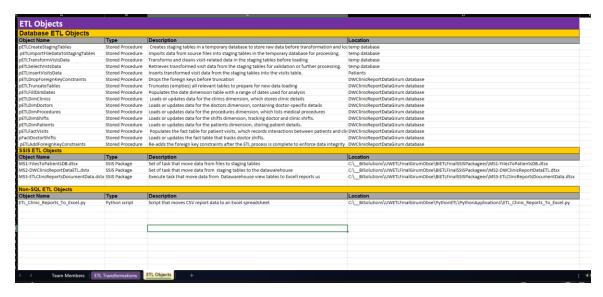


Figure 15, demonstrating the data in DataClinicReport datawarehouse tables

ource DB to Data Warehouse					
ource Data					
	Source Type	Destination	Destination Type	Transformations	
octorsSchedules.dbo.Clinics	Table	DWClinicReportDataGirum.dbo.DimClinics	Table		
A	NA	DWClinicReportDataGirum.dbo.DimClinics.ClinicKey	int	Generated with Identity	
octorsSchedules.dbo.Clinics.ClinicID	int	DWClinicReportDataGirum.dbo.DimClinics.ClinicID	int		
octorsSchedules.dbo.Clinics.ClinicName	nvarchar(100)	DWCIinicReportDataGirum.dbo.DimCIinics.ClinicName	nvarchar(100)		
octorsSchedules.dbo.Clinics.City	nvarchar(100)	DWClinicReportDataGirum.dbo.DimClinics.ClinicCity	nvarchar(100)		
octorsSchedules.dbo.Clinics.State	nvarchar(100)	DWClinicReportDataGirum.dbo.DimClinics.ClinicState	nvarchar(100)		
octorsSchedules.dbo.Clinics.Zip	nvarchar(5)	DWClinicReportDataGirum.dbo.DimClinics.ClinicZip	nvarchar(5)		
DoctorsSchedules].[dbo].[Doctors]	Table	DWClinicReportDataGirum.dbo.DimDoctors	Table		
A	int	DWClinicReportDataGirum.dbo.DimDoctors.DoctorKey	int	Generated with Identity	
octorsSchedules.dbo.Doctors.DoctorID	int	DWClinicReportDataGirum.dbo.DimDoctors.DoctorID	int		
octorsSchedules.dbo.Doctors.FirstName	nvarchar(100)	DWClinicReportDataGirum.dbo.DimDoctors.DoctorFullName	nvarchar(200)	Combine FirstName and LastName and Cast to Type	
octorsSchedules.dbo.Doctors.LastName	nyarchar(100)	DWClinicReportDataGirum.dbo.DimDoctors.DoctorFullName	nvarchar(200)	Combine FirstName and LastName and Cast to Type	
octorsSchedules.dbo.Doctors.EmailAddress	nvarchar(100)	DWClinicReportDataGirum.dbo.DimDoctors.DoctorEmailAddress	nyarchar(100)		
octorsSchedules.dbo.Doctors.Address	nyarchar(100)	NA .	NA	NA .	
octorsSchedules.dbo.Doctors.City	nvarchar(100)	DWClinicReportDataGirum.dbo.DimDoctors.DoctorCity	nvarchar(100)		
octorsSchedules dho Doctors State	nvarchar(100)	DWClinicReportDataGirum.dbo.DimDoctors.DoctorState	nyarchar(100)		
octorsSchedules.dbo.Doctors.Zip	nyarchar(5)	DWCIInicReportDataGirum.dbo.DimDoctors.DoctorZip	nvarchar(5)		
atients.dbo.Patients	Table	DWClinicReportDataGirum.dbo.DimPatients	Table		
A	NA.	DWCIInicReportDataGirum.dbo.DimPatients.PatientKey	int	Generated with Identity	
atients.dbo.Patients	int	DWClinicReportDataGirum.dbo.DimPatients.PatientID	int		
atients dbo Patients Fname	varchar(28)	DWClinicReportDataGirum.dbo.DimPatients.PatientFullName	varchar(100)	Combine FName and LName and Cast to Type	
atients.dbo.Patients.Lname	varchar(29)	DWClinicReportDataGirum.dbo.DimPatients.PatientFullName	varchar(100)	Combine FName and LName and Cast to Type	
atients.dbo.Patients.Email	varchar(100)	NA .	NA.	NA .	
atients dbo Patients Address	Varchar(97)	NA .	NA	NA .	
atients dbo Patients City	varchar(72)	DWClinicReportDataGirum.dbo.DimPatients.PatientCity	varchar(100)		
atients dbo Patients State	varchar(50)	DWClinicReportDataGirum.dbo.DimPatients.PatientState	varchar(100)		
atients.dbo.Patients.ZipCode	int	DWClinicReportDataGirum.dbo.DimPatients.PatientZipCode	int		
A	NA NA	DWClinicReportDataGirum.dbo.DimPatients.StartDate	date	is assigned the current date when a new row is inserted./are managed by the ETL process.	
A.	NA NA	DWCIInicReportDataGirum.dbo.DimPatients.EndDate	date	is updated to the current date when a record is no longer current./are managed by the ETL process.	
A.	NA NA	DWClinicReportDataGirum.dbo.DimPatients.IsCurrent	int	is used as a flag to mark the active record (1 for active, 0 for inactive) are managed by the ETL process.	
Patients].[dbo].[Procedures]	Table	DWCIInicReportDataGirum.dbo.DimProcedures	Table	is used as a ring to mark the active record (1 for active, o for mactive), are managed by the citc process.	
A attents].[abo].[Frocedures]	int	DWClinicReportDataGirum.dbo.DimProcedures.ProcedureKey	int	Generated with Identity	
atients.dbo.Procedures.ID	int	DWCIInicReportDataGirum.dbo.DimProcedures.ProcedureID	int	Generated with identity	
atients.dbo.Procedures.ProcedureName	varchar(100)	DWClinicReportDataGirum.dbo.DimProcedures.ProcedureName	varchar(100)		
atients.dbo.Procedures.ProcedureDesc	varchar(1000)	DWCIInicReportDataGirum.dbo.DimProcedures.ProcedureDesc	varchar(1000)		
atients.dbo.Procedures.ProcedureDesc atients.dbo.Procedures.ProcedureCharge	money	DWCIInicReportDataGirum.dbo.DimProcedures.ProcedureDesc DWCIinicReportDataGirum.dbo.DimProcedures.ProcedureCharge	money		
octorsSchedules.dbo.Shifts	Table	DWCIInicReportDataGirum.dbo.DimProcedures.ProcedureCharge DWCIInicReportDataGirum.dbo.DimShifts	Table		
octorsschedules.dbo.snifts	int	DWClinicReportDataGirum.dbo.DimShifts.ShiftKey	int	Generated with Identity	
A octorsSchedules dbo Shifts	int	DWCIInicReportDataGirum.dbo.DimShifts.ShiftiD	int	Managara and managara	
octorsSchedules.dbo.ShiftsStart	time(0)	DWClinicReportDataGirum.dbo.DimShifts.ShiftStart	time(0)		
octorsSchedules.dbo.ShiftsStart octorsSchedules.dbo.ShiftsEnd					
octorsachedures.ddo.anntSEN0	time(0) Table	DWClinicReportDataGirum.dbo.DimShifts.ShiftEnd DWClinicReportDataGirum.dbo.FactDoctorShifts	time(0) Table		

Figure 16, demonstrating the metadata worksheets



6. Milestone 3: Non-SQL ETL

Process Steps

Creating SQL Views

- Develop two SQL views in the DWClinicReportData database:
 - o One for doctor shifts data.
 - o One for patient visits data.
- Ensure the views contain relevant columns for analytical reporting.
- Validate the views by querying sample data.

Sql code for the views..

Extract, Transform, and Load (ETL) using Python

- Develop a Python script to perform ETL operations:
 - o Extract data from the SQL views.
 - o Transform the data as needed (e.g., date formatting, column renaming).
 - o Load the data into an Excel spreadsheet named ClinicReportsData_.xlsx.
- Store the Excel file in the solution's Reports folder.
- Validate that the data is correctly loaded into the two worksheets (one per view).

Screernshot of the python script for the ETL for Clinic Reports

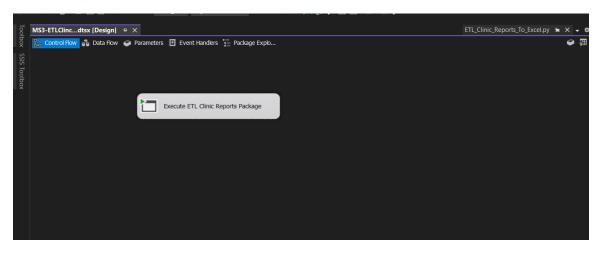
```
**Speny as **Speny as
```

SSIS Package Integration

- Develop an SSIS package (ETLClinicReportsDocumentData.dtsx):
 - o Automate execution of the Python script.
 - o Handle errors and log execution results.
 - o Ensure proper folder and file management for storing output reports.

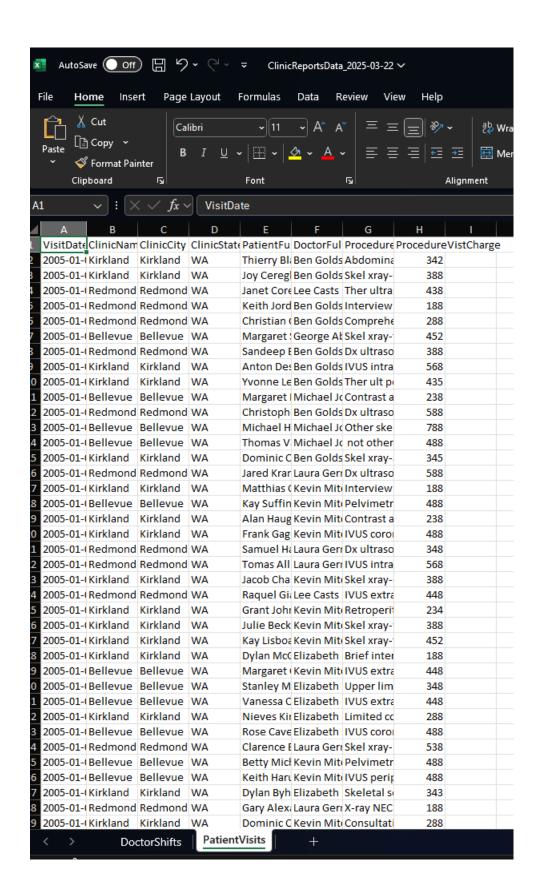
Updating ETL Process Metadata

- Update the Excel metadata spreadsheet:
 - o **Document source, destination, and transformation rules.**
 - o Ensure all tabs are updated as per the ETL workflow.
- Verify the metadata consistency with the implemented ETL process.



Ssis package the executes the python ETL script

Excel worksheet of the clinic reports



7. Milestone 4: Automation, Reports, and Documentation

Process Steps

Creating SQL Reporting Views

- Develop ETL report views to extract data from the ETL log and MSDB jobs tables.
- Ensure the views include relevant columns for tracking ETL execution status, errors, and processing times.
- Validate the views by running queries to confirm correct data extraction and filtering.
- Optimize performance by adding appropriate indexing where necessary.

```
MS3-ETLClincR....dtsx [Design]

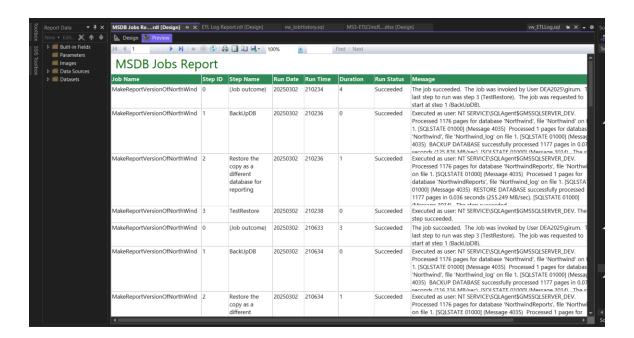
| VW_ETLLog.sq
| V
```

Figure 20, sql script for the views vw_ETL Log and vw_JobHistory

Creating ETL Dashboard Reports with SSRS

Create an ETL report that shows the contents of the ETL log and MSDB jobs table.

- The report should display execution details, success/failure statuses, error messages, and timestamps.
 - Use appropriate filters and sorting options to ensure the report provides useful insights.
- Ensure that your Data Sources and DataSets are properly configured and named according to their respective database objects.



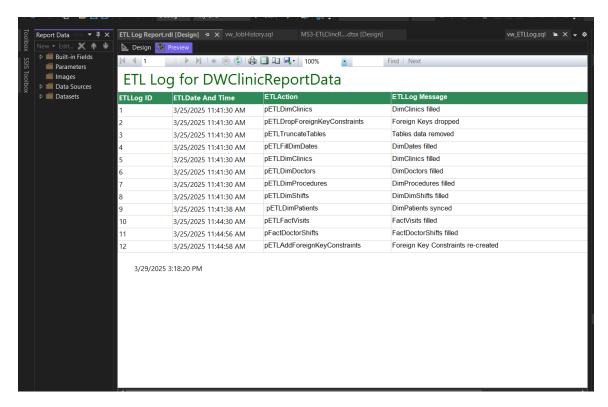


Figure 22, screeshot of the SSRS report for MSDB jobs and ETL Log Report

Summery

The ETL Technical Manual outlines the implementation of an automated ETL solution for a small medical clinic, replacing the current manual CSV-based process with a structured system using SQL Server, SSIS, and Python.

The project involves automating data ingestion, transformation, and loading into a Data Warehouse, improving efficiency and accuracy. Key milestones include file-based ETL for CSV imports, data warehouse ETL for structured reporting, Python-based ETL for Excelbased analytics, and automation using SSIS.

The solution also integrates SQL views, stored procedures, and SSRS dashboards to track ETL performance and errors. Designed for scalability and reliability, this manual serves as a guide for data engineers, ETL developers, and BI analysts involved in data integration projects.