

# INCREMENTAL LOAD/CREATION OF DIMENSION TABLE LOGIC AT MAYO CLINIC

In this particular document we shall be looking at the incremental load logic at Mayo Clinic. This document is being prepared because of the following purposes.

- Data Integrator does not have a dimension creation wizard. Thus using the transforms within the Data Integrator we need to make the required type-1 and type-2 dimensions.
- This document will serve as a template for all new developers to populate the data warehouse and the data mart as part of the PAMDSS project at Mayo Clinic.
- The document is attached with the required templates which shall further help in understanding of the processes. With this document we seek to systematize the development of history, base and the dimension tables within the project.
- Prevent some basic mistakes that developers can make while populating the history and the dimension tables.

Note: Along with this document we shall be attaching/referring to the required *atl* data flows. As part of the understanding we shall be attaching the screen shots of the required *atl* and sending the required *atl*.

## Methodology for populating the Data Warehouse

Scenario for loading the data warehouse: depending upon the project requirements a client within the Data warehouse can have the following tables:

- Base tables
- History tables.

In this document we shall be looking at populating the History tables. The methodology for populating the base tables is simpler and does not require special treatment.

We have history making events and non-History making events. We shall have to track inserts, updates and deletes for both the events. The scenario can be expressed by the use of the following matrix.

<b>Event</b>	<b>Insert</b>	<b>Update</b>	<b>Delete</b>
History making attributes	1	2	3
Non-history making attributes	4	5	6

Also many times we have last update time stamps in the source. We shall be looking at a specific case where we have the time stamp in the source and have to populate the target in such a scenario.

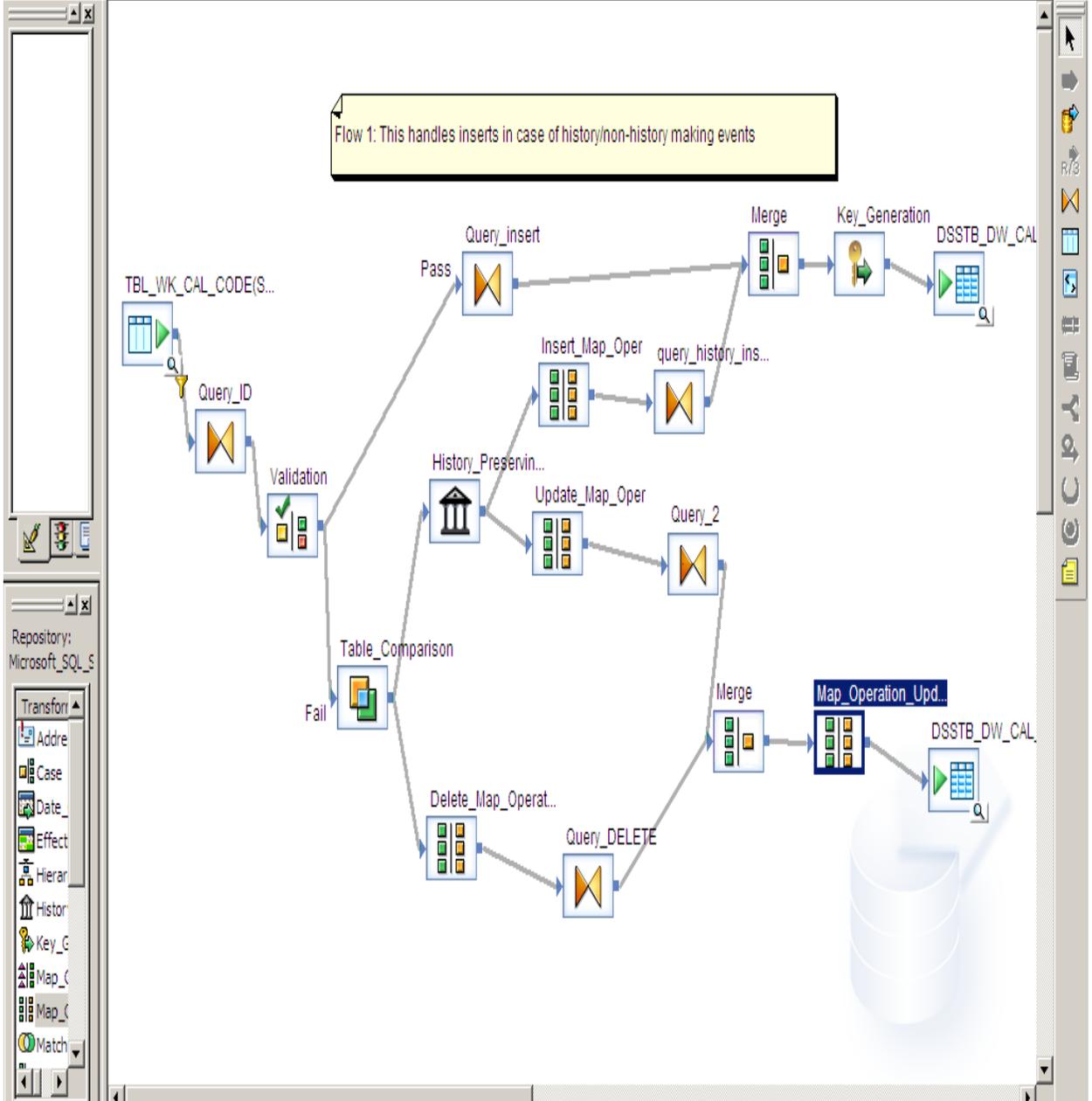
Methodology for handling Inserts: That is in this case we shall be looking at scenarios 1 and 4.

There shall be two kinds of inserts within this scenario.

- Those records whose business key is not present in the data warehouse.
- Those history making records that have been updated. We shall generate an insert row for each one of them.

**Methodology for handling those records whose business key is not present in the Data warehouse**

For this purpose we propose that you follow activities within Flow 1. In the following diagram Flow1 includes Query\_ID, Validation, Query\_Insert, Merge and Surrogate key generation.



Steps in the process are as following:

1. Lookup for the business key in the data warehouse table.
2. Validate whether the Business key is Null or not. Within the Validation transform put the “Business key is NULL” as the custom condition.
3. In case Business Key is NULL then it is straight a case of an insert.
4. Set the Insert Time Stamp, Update Time Stamp, Effective From (History Start) Time Stamp as today’s date [Query\_Insert].
5. Set the Effective through(History End time stamp) as the end of time.

*Note: This shall handle Insert operation for all non-history making events also.*

Handling Inserts from History Making Records: for the records that are to be updated we shall be handling separately within the updates section. However for the new row to be inserted after the history preserving transform, we perform the following operations.

- Map the Insert row as normal (discard all the other row types) through the use of Map Transform
- Merge the output with the Inserts from Flow1.
- Generate a surrogate key after union operation.

Handling Updates for History Making attributes (scenario 3): In this particular case we have the following scenario:

- The Lookup value of the business key(from Query\_ID) is not NULL.
- We have History Start and History End Time stamps in the target table.
- We know the attributes on which we propose to preserve history.

Then the combination of activities used shall be as following:

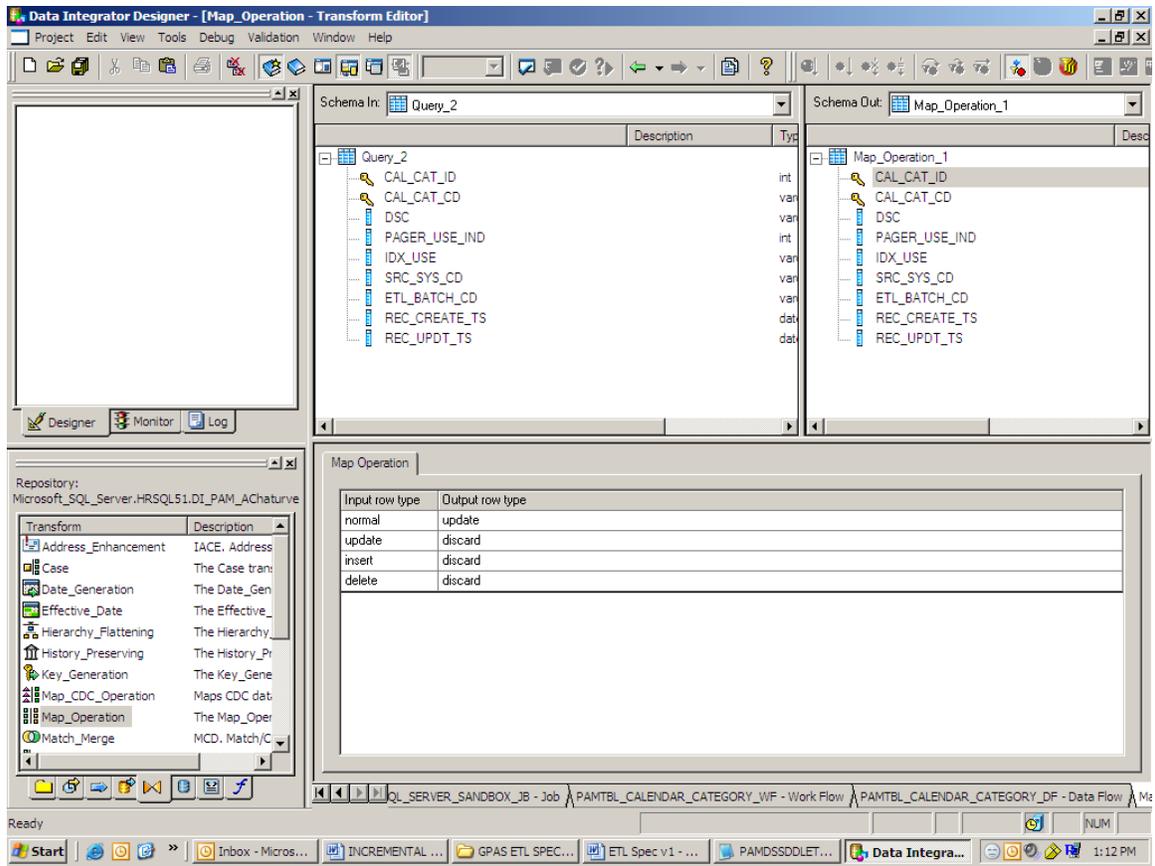
- Table Comparison: this shall help determine all the required updates within the target table. All the output rows from this transform shall be marked as ‘U’. The following care should be taken when we use this transform:
  - The Compare column should not have the Target table Load Time Stamps and any of the primary key columns.

- The compare column should have only the history tracking attributes from the target table.
- History Preserving transform: The History preserve transform shall have input rows marked as 'Update'. Within the History preserve transform we need to take the following care:
  - The history preserve columns are the same as in Table Comparison.
  - If there is a row status in the target table then its *Reset Status* should be changed.
- Map the update rows: this is done using the MAP transform and only marking the Update rows output as Normal. The other row types should be discarded.

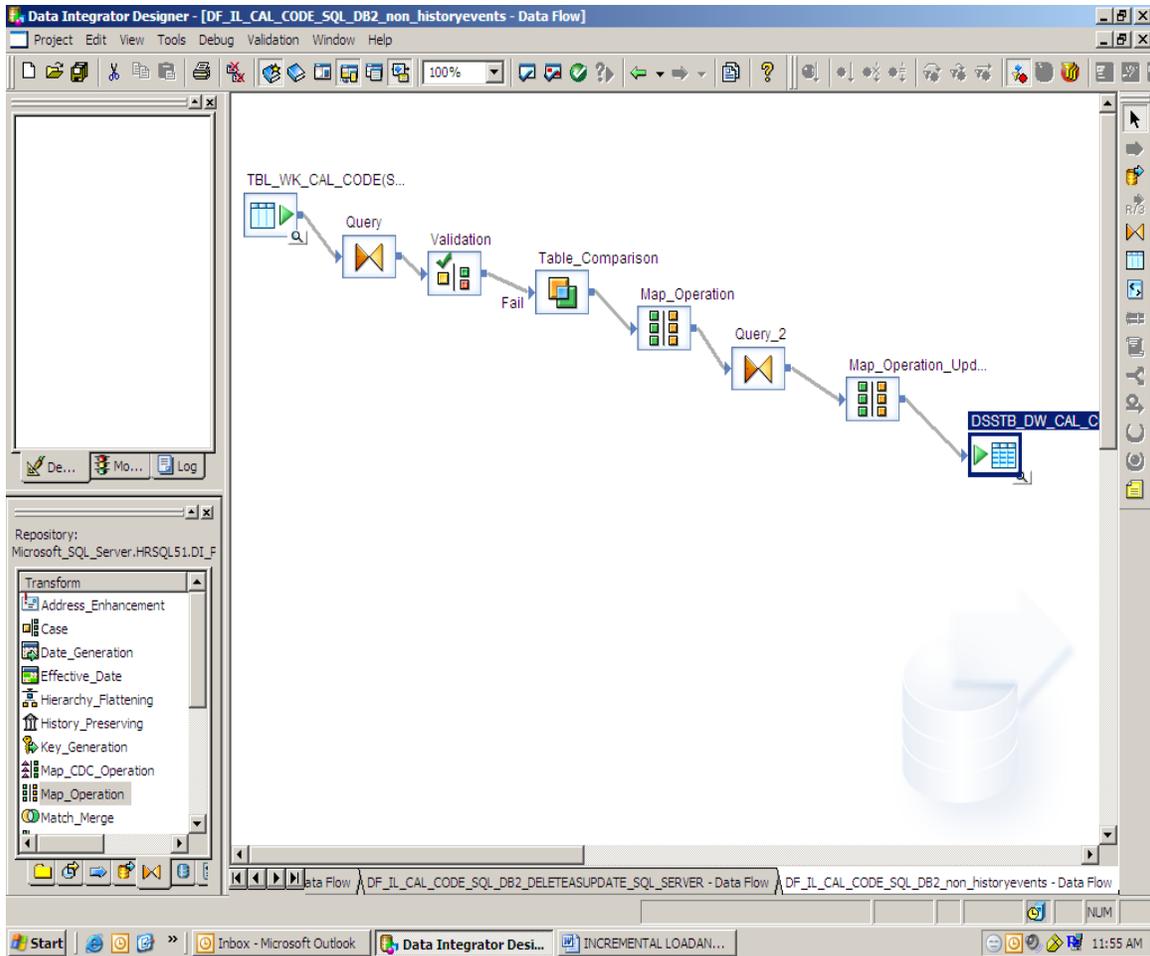
Handling Deletes: within Data Integrator we have an option for marking Deletes as Update rows within the History Preserve transform. Our experience with the same has not been very convincing. The alternate methodology is as following:

- In the Table Comparison, use the option Preserve deleted rows.
- Map the deleted rows using MAP transform(discard all input types except the deletes).
- Set the last update time stamp and the History end date as today's date.
- Set the row status as 'delete'.
- Merge it with the update rows.

**Explanation for the transform Map\_Operation\_Update:** It is important to consider why we need to add one last Map transform before we load the data into the target table. For case of *Normal*(we converted the update rows into Normal rows) rows mark the respective row as *Update*(as given in the following diagram). The reason for doing the given step is that in case the rows are marked as normal then they shall be inserted as a separate row in the target table. In case we have primary key constraints in the target it shall then result in an error.



Handling Updates for the Non-history making attributes (Scenario 5): Unfortunately we cannot handle this case within the same data flow. For this we designed another data flow whose implementation shall follow the execution of the previous data flow. In this case we simply track the rows flagged as *Updates* from the Table Comparison Transform in the following manner:



In the Query\_2 step we shall set the Last Update Time Stamp to today's date.

(Note we should take the following care while implementing the above steps)

- In the Compare column category of Table Comparison include only the non-History making attributes from the target table. Do not include the target table load Time Stamps and the primary key columns in the Compare column category.
- In the last Map operation Map\_Operation\_Update set the Input *Normal* rows as output type *update*.

**Input Tables have a Last Update Time stamp:** In case the input tables have a last update time stamp then we can limit the number of rows to be scanned. This is done with the help of the following steps:

- Generate a script before the dataflow in which you capture the Maximum value of the **Last Update Time stamp from the target table**(see the marked rows in red in the attachment below). This can be done by writing a simple script as following:

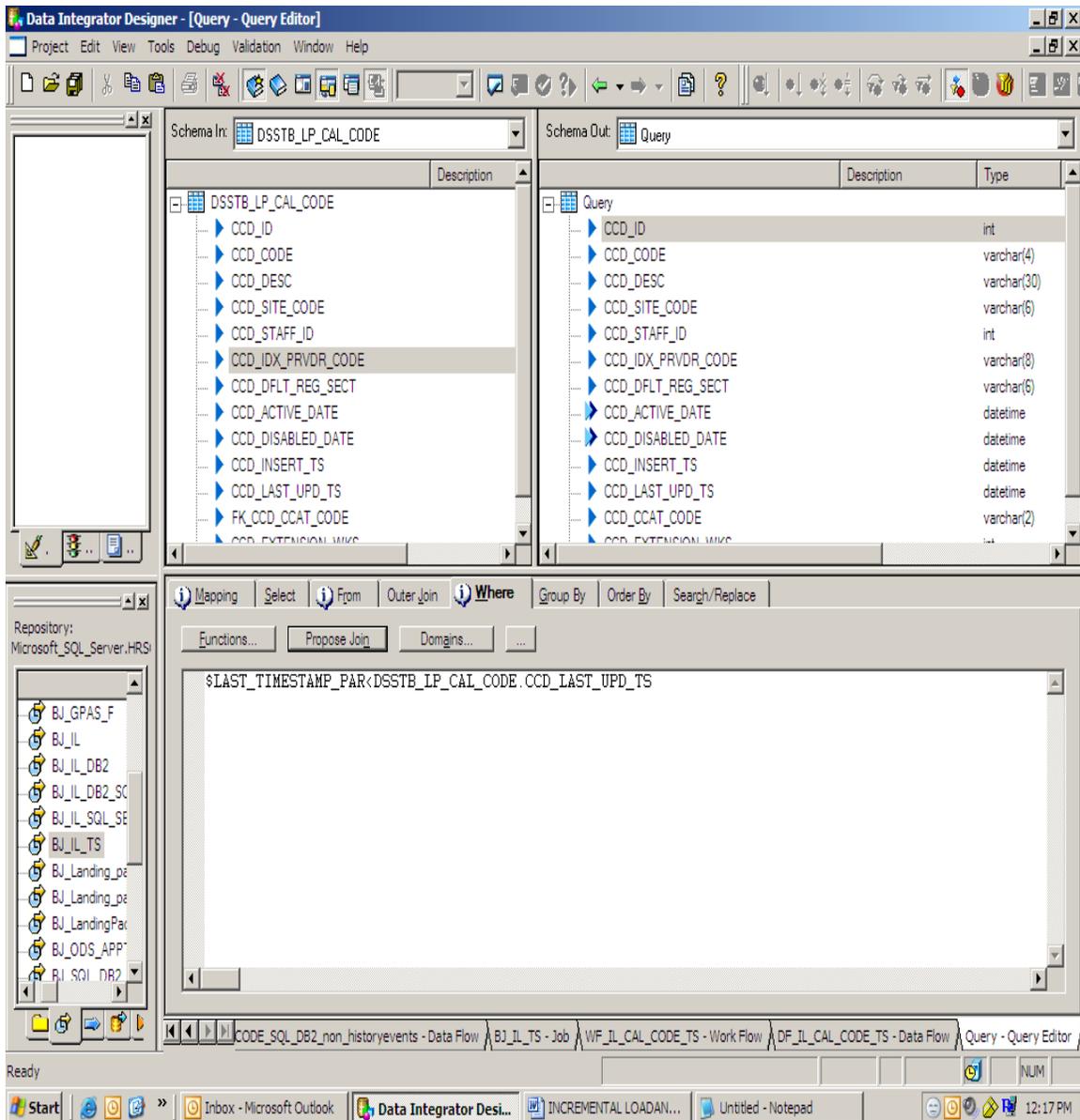
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$FULLUPLOAD='NO';
IF ($FULLUPLOAD='YES')
begin
    print( 'The process of full upload is set to begin');
    sql( 'SQL_SERVER_SANDBOX', 'TRUNCATE TABLE DSSTB_DW_CAL_CODE_TS');
    sql(
        'SQL_SERVER_SANDBOX', 'INSERT
        INTO
        DSSTB_DW_CAL_CODE_TS(CCD_DW_ID,CCD_DW_ROW_STATUS,CCD_DW_INSERT_TS,
        CCD_DW_LAST_UPD_TS,CCD_ID,CCD_CODE,CCD_DESC,CCD_SITE_CODE,CCD_STAF
        F_ID,CCD_IDX_PRVDR_CODE,CCD_DFLT_REG_SECT,CCD_ACTIVE_DATE,CCD_DISAB
        LED_DATE,CCD_INSERT_TS,CCD_LAST_UPD_TS,CCD_CCAT_CODE,CCD_EXTENSION_
        WKS)
        VALUES(0,\',\'\'1900-01-01
        00:00:00.000\',\'1900-01-01
        00:00:00.000\',0,\MA\',\';\';0,\';\';\'1900-01-01
        00:00:00.000\',\'1900-01-01 00:00:00.000\',\'1900-01-01 00:00:00.000\',\';0)');
    $Last_Timestamp_var='1900-01-01 00:00:00.000';
end
else
begin
    $Last_Timestamp_var=SQL('SQL_SERVER_SANDBOX','SELECT
    MAX(CCD_LAST_UPD_TS) FROM DSSTB_DW_CAL_CODE_TS');
    print('the process of incremental load has started');
end

```

Thereafter we do the following steps:

- Declare the variable as a parameter.
- Pass the parameter into the dataflow.
- In the 'where' clause (as given in the attachment below) select the time stamps from the **source table with values greater than the parameter value.**

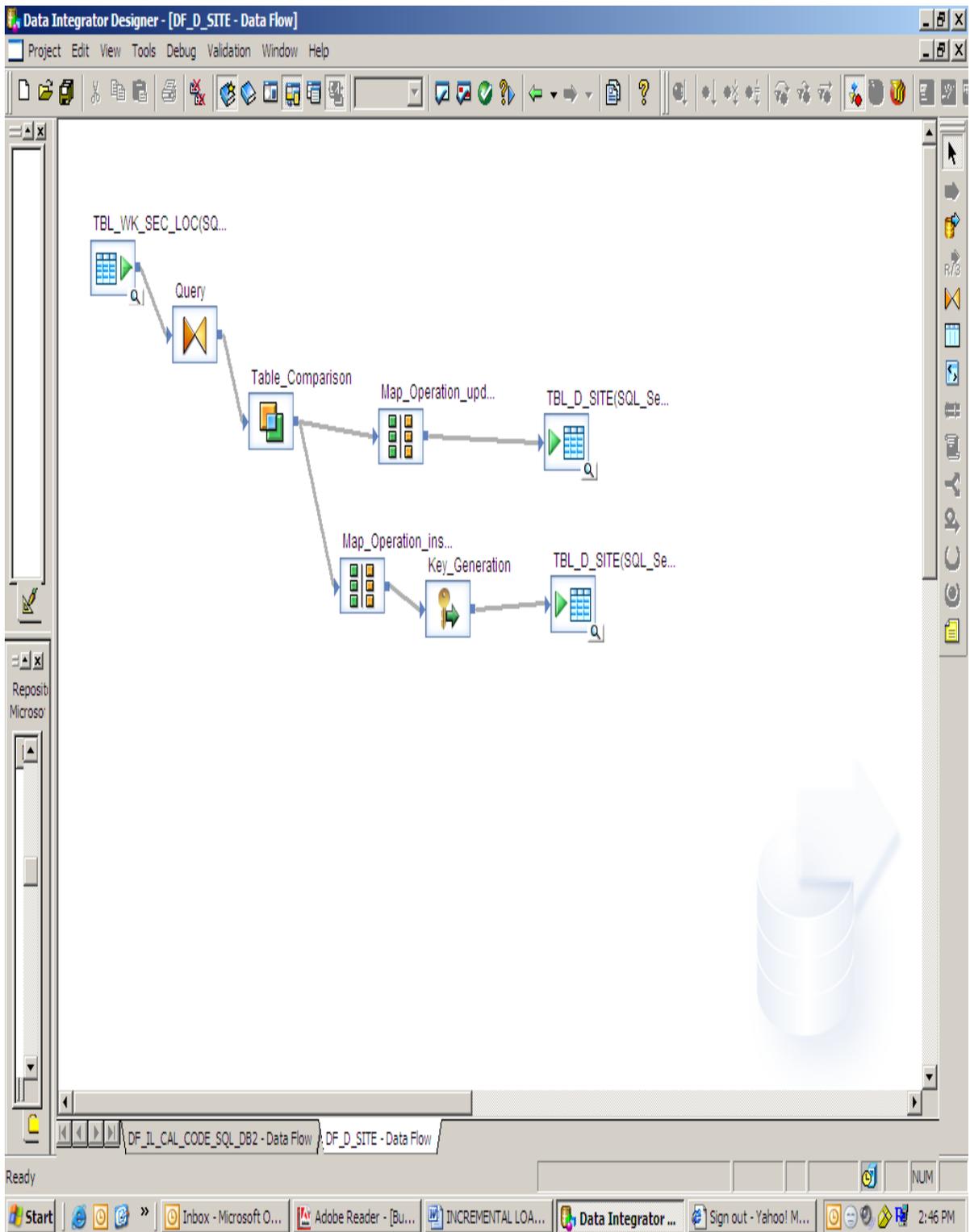


Dimension Tables: In this case we shall look at both Type-1 and Type-2 dimension tables.

How to work with Type-1 Dimension tables: In this case we wish to carry out the following steps:

- In case of an update keep the latest record. Do not increment the dimension key.
- However in case of an insert we wish to increment the dimension key.

We carry the above using the following steps:



The logic: the logic for the above data flow is as following.

From the source tables we shall get records marked as Insert(new records) and updates(old records changed). In case the record is an update then load it directly onto the target table. This is done by passing it through a map transform(*the output type for the Update being 'Update' with other rows types being 'discarded'*). However in case it is an insert, then we generate the surrogate key.

Methodology for populating type-2 Dimension Tables: The methodology for populating the type-2 dimension tables is the same as populating the history tables with history making attributes. The sequence of operation and the logic are as following:

