

Community Health Status Indicator Information Management Solution



Team Number: 12

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Course: INFO 7290

Course Name: Data Warehousing and Business Intelligence

Guided By: Prof. Vincent Lattuada

Objectives:

- To find multiple sources of data to form a substantial SOR required for warehousing.
- To provide a brief walk through of building a data warehouse that will house Health-centric demographic data granularized at the 'county' level.
- To implement advanced Data- warehousing techniques while answering questions with a Business perspective.
- To provide an overview of the Data Flow Process.
- Serve as Version - log for future references

Version History:

| Date | Submission | Description |
|------------|--------------------|---|
| 11/05/2020 | Project Proposal-1 | First Draft |
| 11/15/2020 | Project Proposal-2 | Modified the First draft with the changes instructed by professor |
| 12/04/2020 | Project Design Doc | Added some more design related details like data flow, fields, olap cubes, SSIS snapshots, Visualizations |
| 12/16/2020 | Final Draft | Updated dimensional model, and listed additional features |

Table of Contents:**1. Description of the data to be stored and moved into the analytics system**

- Data Source Description
- Joining Data sets
- Data Transformations
- Dataset Dimensions
- Field/Type Details

2. High level data flow

- Data flow
- Dimensional Model
- Incremental Loads
- Lookup Tables
- Data marts/ OLAP cubes

3. Analytics Design

- Reporting tools and visualizations

4. Analytics outcomes

- Interesting facts

1. Description of the data to be stored and moved into the analytics system

1.1. Data Source Description

Community Health Status Indicators (CHSI) to combat obesity, heart disease, and cancer are major components of the Community Health Data Initiative.

1. Leading Cause of Death: This dataset contains details leading cause of deaths among various American counties, it gives detailed information regarding the age group and cause of death.

https://data.world/health/chsi-to-combat-obesity/workspace/file?filename=chsi_dataset%2FLEADINGCAUSESOFDEATH.csv

2. Preventive Services Used across the counties: This data source gives us the idea about various preventive measures in terms of vaccine taken across the counties. It includes data of vaccines like Hepatitis, Measles, Syphilis, Flu Vaccine. It also gives a bifurcation among the age groups and vaccines taken by them. This data can be joined with the previous set to analyse preventive services vs deaths.

https://data.world/health/chsi-to-combat-obesity/workspace/file?filename=chsi_dataset%2FPREVENTIVESERVICESUSE.csv

3. Risk Factors and Access to Care: This data source adds a new aspect to the analysis plan. It consists of data regarding the risk factor among counties like resident with high blood pressure, diabetes, obesity. The data source also contains data regarding the accessibility to care like elderly Medicare, insurance, disability care, dentists, etc.

https://data.world/health/chsi-to-combat-obesity/workspace/file?filename=chsi_dataset%2FRISEFACTORSANDACCESSTOCARE.csv

4. Demographics: This data source is our base source as it contains information regarding the population density of each county. This dataset also contains percentiles in respective to peer counties.

https://data.world/health/chsi-to-combat-obesity/workspace/file?filename=chsi_dataset%2FDEMOGRAPHICS.csv

1.2. Joining Data Sources

The datasets Leading Cause of Death and Preventive Services Used across the counties, Risk Factors and Access to Care, can be joined by a composite primary key formed by the State_FIPS_Code and County_FIPS_Code. We are forming a composite primary key here as

the State_FIPS_Code and County_FIPS_Code contains repeated values but if used together they form a unique key for each record.

1.3.Data Transformations

Our dataset contains following indicators:

| | data_value | description |
|----|------------------------|--|
| 1 | -9999 | Indicate N.A. value from the source data for the Unemployed column on the VUNERABLEPOPSANDEN |
| 2 | -2222 or -2222.2 or -2 | nda, no data available, see Data Notes document for details |
| 3 | -1111.1 or -1111 or -1 | nrf, no report, see Data Notes document for details |
| 4 | 1 | Represent 'No' in the indicator columns |
| 5 | 2 | Represent 'Yes' in the indicator columns |
| 6 | 3 | Represent 'Favorable to peers' in the indicator columns |
| 7 | 4 | Represent 'Unfavorable to peers' in the indicator columns |
| 8 | 5 | Represent 'Favorable to peers and favorable the U.S. Rate' in the indicator columns |
| 9 | 6 | Represent 'Favorable to peers and unfavorable the U.S. Rate' in the indicator columns |
| 10 | 7 | Represent 'Unfavorable to peers and favorable the U.S. Rate' in the indicator columns |
| 11 | 8 | Represent 'Unfavorable to peers and unfavorable the U.S. Rate' in the indicator columns |
| 12 | -9998.9 | Indicate no objective for the Healthy People 2010 Target data |

Fig : 1- Data transformations

The dataset contains numeric value in place of indicators like “Yes” or “No” for example 1 represents “No” 2 represents “Yes”. We plan to transform those numeric values to “Yes”, “No”, “Favorable”, “Unfavorable” so we can use them as categories during analysis.

The dataset has been built keeping in mind the various missing values encountered in the real world. Some of these values look like extreme integers such as ‘-9999’, ‘-1111’, etc. For simplicity we will transform these values into ‘NULL’ values.

As shown above we need to assign a consistent value to these indicators because if they are not transformed they may produce erroneous results when aggregated leading to a false analysis.

1.4 Dataset Dimensions: These datasets have following dimensions.

- Age
- Race
- State and County
- Disease
- Cases

- Testing
- Risk factors

1.5 Fields and DataTypes

Dataset: Leading Causes of Death

Corresponding Staging Table: Leading_Causes_Staging

Corresponding Dimension Table: Dim.Leading_Causes

| IN DATASET | FOR STAGING | DATATYPE | SIGNIFICANCE |
|------------------|------------------|----------|---|
| STATE_FIPS_CODE | STATE_FIPS_CODE | INT | Two-digit state identifier |
| COUNTY_FIPS_CODE | COUNTY_FIPS_CODE | INT | Three-digit county identifier |
| CHSI_STATE_NAME | STATE_NAME | VARCHAR | Name of State |
| CHSI_COUNTY_NAME | COUNTY_NAME | VARCHAR | Name of county |
| CHSI_STATE_ABBR | STATE_ABBR | CHAR(2) | Two-character abbreviation for state name |
| B_WH_CANCER | CANCER_WH_A1 | INT | County data, ages 1-14, cancer, White |
| B-BL-CANCER | CANCER_BL_A1 | INT | County data, ages 1-14, cancer, Black |
| B_Ot_Cancer | CANCER_OT_A1 | INT | County data, ages 1-14, cancer, Other |

| | | | |
|-------------|--------------|-----|--|
| C_Wh_Cancer | CANCER_WH_A2 | INT | County data, ages 15-24, cancer, White |
| C_BL_CANCER | CANCER_BL_A2 | INT | County data, ages 15-24, cancer, Black |
| C-OT_CANCER | CANCER_OT_A2 | INT | County data, ages 15-24, cancer, Other |
| D-WH-CANCER | CANCER_WH_A3 | INT | County data, ages 25-44, cancer, White |
| D-BL-CANCER | CANCER_BL_A3 | INT | County data, ages 25-44, cancer, Black |
| D_ot-CANCER | CANCER_OT_A3 | INT | County data, ages 25-44, cancer, other |
| E-WH-CANCER | CANCER_WH_A4 | INT | County data, ages 45-64, cancer, white |
| E_BL_CANCER | CANCER_BL_A4 | INT | County data, ages 45-64, cancer, black |
| E_Ot_Cancer | CANCER_OT_A4 | INT | County data, ages 45-64, cancer, other |
| F_WH_CANCER | CANCER_WH_A5 | INT | County data, ages 65+, cancer, white |
| F_BL_CANCER | CANCER_BL_A5 | INT | County data, ages 65+, cancer, black |

| | | | |
|---------------|--------------|-----|---|
| F_Ot_Cancer | CANCER_OT_A5 | INT | County data, ages 65+, cancer, white |
| D_Wh_HeartDis | HD_WH_A3 | INT | County data, ages 25-44, heart disease, White |
| D_Bl_HeartDis | HD_Bl_A3 | INT | County data, ages 25-44, heart disease, Black |
| D_Ot_HeartDis | HD_OT_A3 | INT | County data, ages 25-44, heart disease, Other |
| E_Wh_HeartDis | HD_WH_A4 | INT | County data, ages 45-64, heart disease, White |
| E_Bl_HeartDis | HD_Bl_A4 | INT | County data, ages 45-64, heart disease, Black |
| E_Ot_HeartDis | HD_OT_A4 | INT | County data, ages 45-64, heart disease, other |
| F_Wh_HeartDis | HD_WH_A5 | INT | County data, ages 65+, heart disease, White |
| F_Bl_HeartDis | HD_Bl_A5 | INT | County data, ages 65+, heart disease, Black |
| F_OT_HeartDis | HD_OT_A5 | INT | County data, ages 65+, heart disease, Other |

| | | | |
|---------------|-----------|-----|--|
| D_Wh_HIV | HIV_WH_A3 | INT | County data, ages 25-44, hiv/aids, White |
| D_Bl_HIV | HIV_Bl_A3 | INT | County data, ages 25-44, hiv/aids, Black |
| D_Ot_HIV | HIV_OT_A3 | INT | County data, ages 25-44, hiv/aids, Other |
| C_Wh_Suicide | SUI_WH_A2 | INT | County data, ages 15-24, suicide, White |
| C_Bl_Suicide | SUI_Bl_A2 | INT | County data, ages 15-24, suicide, Black |
| C_Ot_Suicide | SUI_OT_A2 | INT | County data, ages 15-24, suicide, Other |
| D_Wh_Suicide | SUI_WH_A3 | INT | County data, ages 25-44, suicide, White |
| D_Bl_Suicide | SUI_Bl_A3 | INT | County data, ages 25-44, suicide, Black |
| D_Ot_Suicide | SUI_OT_A3 | INT | County data, ages 25-44, suicide, Other |
| A_Wh_BirthDef | BD_WH_A | INT | County data, under age 1, birth defects, White |

| | | | |
|---------------|---------|-----|--|
| A_Bl_BirthDef | BD_Bl_A | INT | County data, under age 1, birth defects, Black |
| A_Ot_BirthDef | BD_OT_A | INT | County data, under age 1, birth defects, Other |

Dataset: Risk Factors And Access To Care

Corresponding Staging Table: stg_Riskfactors_Access_Care

Corresponding Dimension Table: Dim.RiskFactors

| IN DATASET | FOR STAGING | DATATYPE | SIGNIFICANCE |
|---------------------|---------------------|----------|-----------------------------------|
| STATE_FIPS_CODE | STATE_FIPS_CODE | INT | Two-digit state identifier |
| COUNTY_FIPS_CODE | COUNTY_FIPS_CODE | INT | Three-digit county identifier |
| NO_EXERCISE | NO_EXERCISE | INT | Deaths due to No Exercise |
| OBESITY | OBESITY | VARCHAR | Deaths due to Obesity |
| HIGH_BLOOD_PRESSURE | HIGH_BLOOD_PRESSURE | INT | Deaths due to High Blood Pressure |
| SMOKER | SMOKER | INT | Deaths due to smoking |

| | | | |
|-------------------|-------------------|-----|---|
| DIABETES | DIABETES | INT | Deaths due to diabetes |
| UNINSURED | UNINSURED | INT | Deaths of the Uninsured |
| ELDERLY_MEDICARE | ELDERLY_MEDICARE | INT | Deaths of the Elderly who had Medicare |
| DISABLED_MEDICARE | DISABLED_MEDICARE | INT | Deaths of the Disabled who had Medicare |

Dataset: Demographics

Corresponding Staging Table: Stg_Demographics

Corresponding Dimension Table: Dim.Demographics

| IN DATASET | FOR STAGING | DATATYPE | SIGNIFICANCE |
|------------------|------------------|----------|---|
| STATE_FIPS_CODE | STATE_FIPS_CODE | INT | Two-digit state identifier |
| COUNTY_FIPS_CODE | COUNTY_FIPS_CODE | INT | Three-digit county identifier |
| CHSI_STATE_NAME | STATE_NAME | VARCHAR | Name of State |
| CHSI_COUNTY_NAME | COUNTY_NAME | VARCHAR | Name of county |
| CHSI_STATE_ABBR | STATE_ABBR | CHAR(2) | Two-character abbreviation for state name |
| Number_Counties | n_counties | INT | Number of Peer Counties |

| | | | |
|--------------------|--------------------|---------|---|
| Population_Size | population | INT | County data, population size |
| Population_Density | population_density | INT | 'County data, population density (people per square mile) |
| Poverty | poverty | DECIMAL | County data, individuals living below poverty level |
| Age_19_Under | under_19 | INT | County data, population under age 19 |
| Age_19_64 | between_19_64 | DECIMAL | County data, population age 19-64 |
| Age_65_84 | between_65_84 | DECIMAL | County data, population age 65-84 |
| Age_85_and_Over | above_85 | DECIMAL | County data, population age 85+ |
| White | white | DECIMAL | County data, White |
| Black | black | DECIMAL | County data, Black |
| Native_American | native | DECIMAL | County data, American Indian |
| Asian | asian | DECIMAL | County data, Asian/Pacific Islander |
| Hispanic | hispanic | DECIMAL | County data, Hispanic origin |

Dataset: Preventive Services Use

Corresponding Staging Table: Preventive_Services_Use

Corresponding Dimension Table: Dim_Preventive_Services

| IN DATASET | FOR STAGING | DATATYPE | SIGNIFICANCE |
|------------------|----------------------------|----------|--|
| STATE_FIPS_CODE | STATE_FIPS_CODE | INT | Two-digit state identifier |
| COUNTY_FIPS_CODE | COUNTY_FIPS_CODE | INT | Three-digit county identifier |
| CHSI_STATE_NAME | STATE_NAME | VARCHAR | Name of State |
| CHSI_COUNTY_NAME | County_Name | VARCHAR | Name of county |
| CHSI_STATE_ABBR | STATE_ABBR | CHAR(2) | Two-character abbreviation for state name |
| FlueB_Rpt | InfluenzaeB_reported_cases | INT | County data, Haemophilus Influenzae B reported cases |
| FlueB_Exp | InfluenzaeB_Expected_cases | INT | County data, Haemophilus Influenzae B expected cases |
| HepA_Rpt | HepatitisA_reported_Cases | INT | County data, Hepatitis A reported cases |

| | | | |
|--------------|--|-----|---|
| HepA_Exp | HepatitisA_Expected_Cases | INT | County data, Hepatitis A expected cases |
| HepB_Rpt | HepatitisB_reported_Cases | INT | County data, Hepatitis B reported cases |
| HepB_Exp | HepatitisB_Expected_Cases | INT | County data, Hepatitis B expected cases |
| Meas_Rpt | Measles_reported_Cases | INT | County data, Measles reported cases |
| Meas_Exp | Measles_Expected_Cases | INT | County data, Measles expected cases |
| Pert_Rpt | Pertussis_reported_cases | INT | County data, Pertussis reported cases |
| Pert_Exp | Pertussis_Expected_cases | INT | County data, Pertussis expected cases |
| CRS_Rpt | Congenital_Rubella_Syndrome reported cases | INT | County data, Congenital Rubella Syndrome reported cases |
| CRS_Exp | Congenital_Rubella_Syndrome_expected_cases | INT | County data, Congenital Rubella Syndrome expected cases |
| Syphilis_Rpt | Syphilis_reported_cases | INT | County data, Syphilis reported cases |

| | | | |
|--------------|---------------------------|-----|--|
| Syphilis_Exp | Syphilis_Expected_cases | INT | County data, Syphilis expected cases |
| ID_Time_Span | Time_period | INT | Time period of reported data for infectious disease cases in the preventive services use |
| Pap_Smear | pap_smears_tests_count | INT | County data, pap smears, age 18+ |
| Mammogram | mammography_tests_count | INT | County data, mammography , age 50+ |
| Proctoscopy | Sigmoidoscopy_tests_count | INT | County data, sigmoidoscopy , age 50+ |
| Pneumo_Vax | pneumonia_vaccine_count | INT | County data, pneumonia vaccine, age 65+ |
| Flu_Vac | flu_vaccine_count | INT | County data, flu vaccine, age 65+ |

2.1.Dataflow Process: A high level overview:

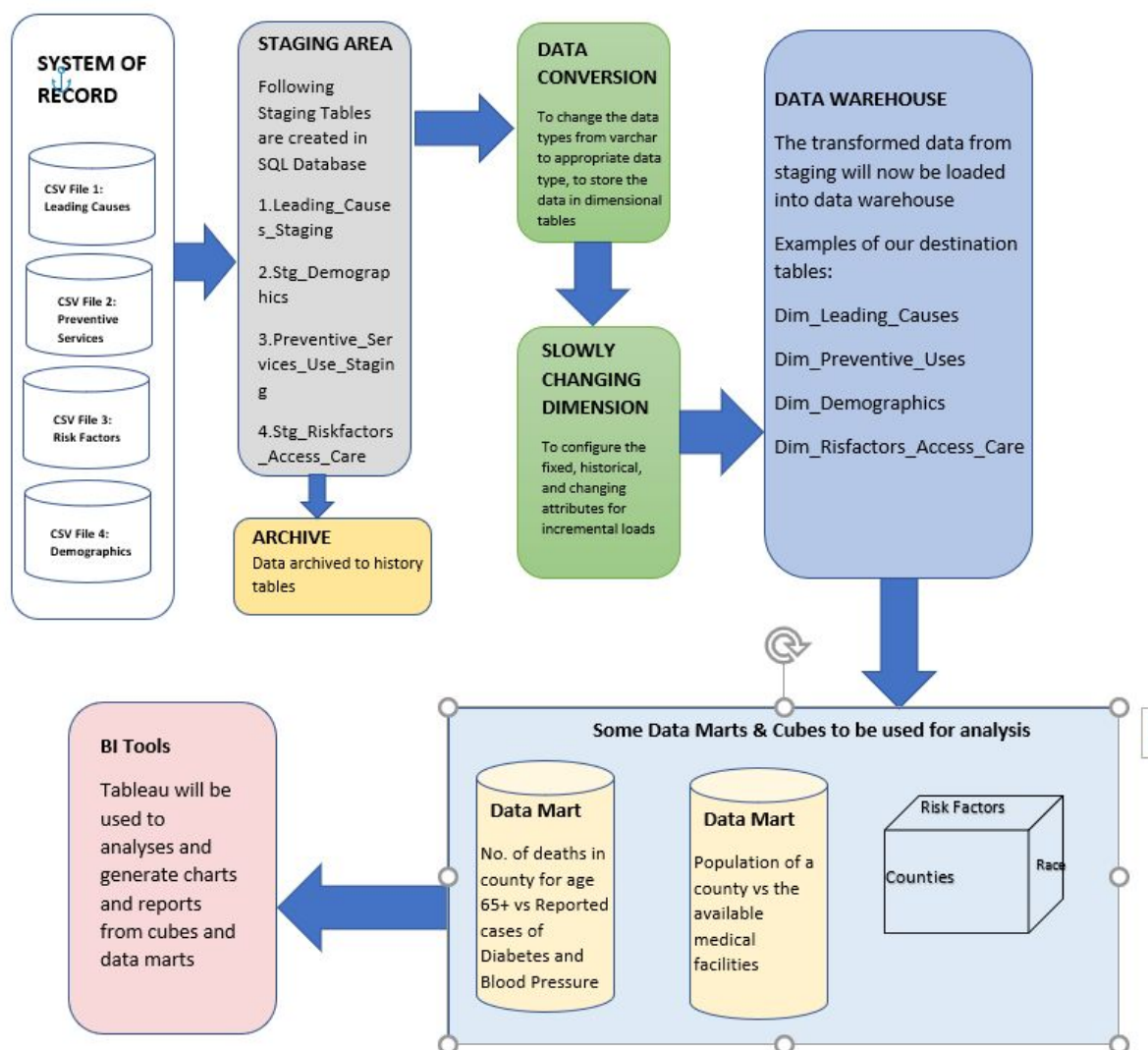


Fig : 2- Data Flow overview

2.2 Dimensional Model

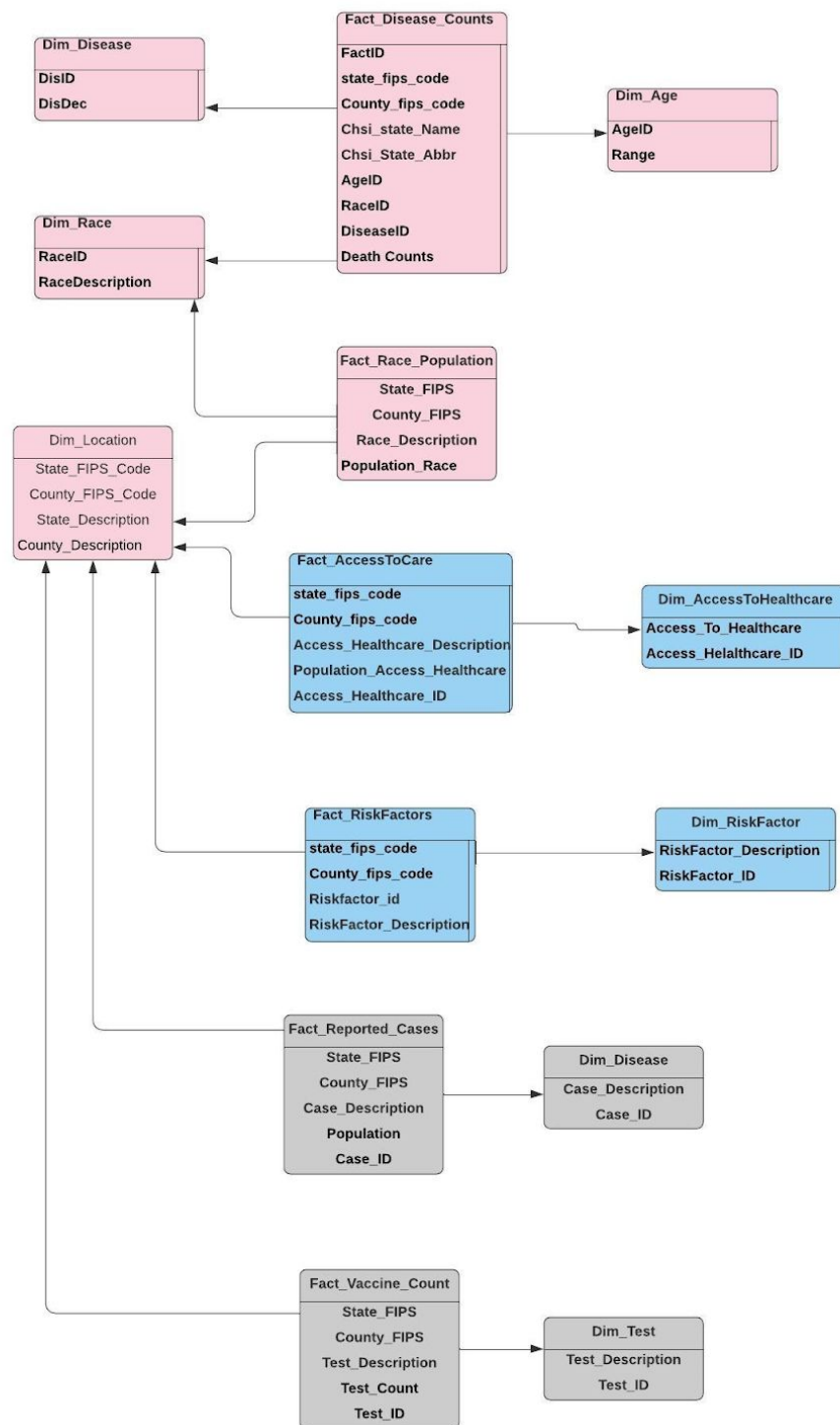


Fig : 3- Dimensional Model

2.3 Incremental loads (slowly changing dimensions)

For the first time loading the dataset can be loaded through SSIS to a staging area and then to the data warehouse. In this process we configure our SSIS package by mapping the input source fields to the destination fields. The initial SSIS package will flow as following:



Fig : 4- Data Flow Simplified

Monthly Loads: Considering the nature of our data sources the data we receive every month should update the existing data or add new records based on the changing field. We have divided all our dimension attributes in three categories namely : fixed, changing and historical. The value of fixed attributes cannot be manipulated in any case eg: County Code, State Code. The changing attributes can be overwritten and we can optionally maintain their prior values. The historical attributes if changed need to be added as an entirely new record, and flagging the old records as expired.

| Dimension Columns | Change Type |
|-----------------------------|----------------------|
| Congenital_Rubella_expected | Changing attribute |
| Congenital_Rubella_reported | Changing attribute |
| County_Name | Fixed attribute |
| flu_vaccine | Historical attribute |
| HepatitisA_Expected_Cases | Changing attribute |
| HepatitisA_reported_Cases | Changing attribute |
| HepatitisB_Expected_Cases | Changing attribute |
| HepatitisB_reported_Cases | Changing attribute |
| InfluenzaeB_Expected_cases | Changing attribute |
| InfluenzaeB_reported_cases | Changing attribute |
| mammography_tests | Changing attribute |
| Measles_Expected_cases | Changing attribute |
| Measles_reported_cases | Changing attribute |
| pap_smears_tests | Changing attribute |
| Pertussis_Expected_cases | Changing attribute |
| Pertussis_reported_cases | Changing attribute |
| pneumonia_vaccine | Changing attribute |
| sigmoidoscopy_tests | Changing attribute |
| State_Abbr | Fixed attribute |
| State_Name | Fixed attribute |
| Syphilis_expected | Changing attribute |
| Syphilis_reported | Changing attribute |
| Time_period | Changing attribute |

Fig 5: Slowly Changing Dimension Object from SSIS

2.4 Lookup Tables

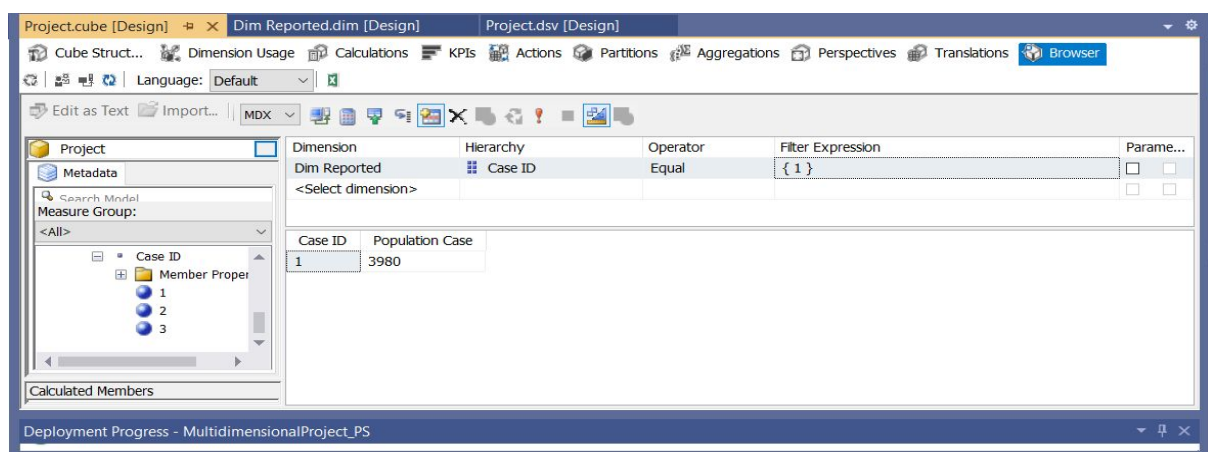
- Lookup tables are used as reference points for decisions regarding correct or incorrect data.
- Lookup tables can be used to check if the data in categorical fields match the listed categories.
- Lookup table for state abbreviation holds the abbreviation for 50 US states, and if data has an abbreviation that does not match then that data is sent to the error table.
- Lookup table for the race field holds values like white, black, hispanic, etc and if there is a race value which does not match it is directed to the error table.

2.5 OLAP Cubes

Keeping reporting and analysis in mind the data can be franchised into different data marts. This is done for snappy reporting and ease of access to relevant data. For e.g. the team that looks after risk factor analysis does not require data related to preventive measures.

A data mart will be created where it will aggregate the data of all the counties of an individual state so we can perform state wise analysis related to the leading cause of death. A sample use case - deaths caused by cancer in NewYork compared to deaths caused by Cancer in Florida.

The other data mart will consist of data comparing the deaths and population data at a county level granularity. This can be done by pulling relevant fields and aggregating to find effective death rate, poverty level, population density, etc. This provides the BI team to look at the overall status of the country and pin-point as to which areas require assistance in terms of medicare and other resources.



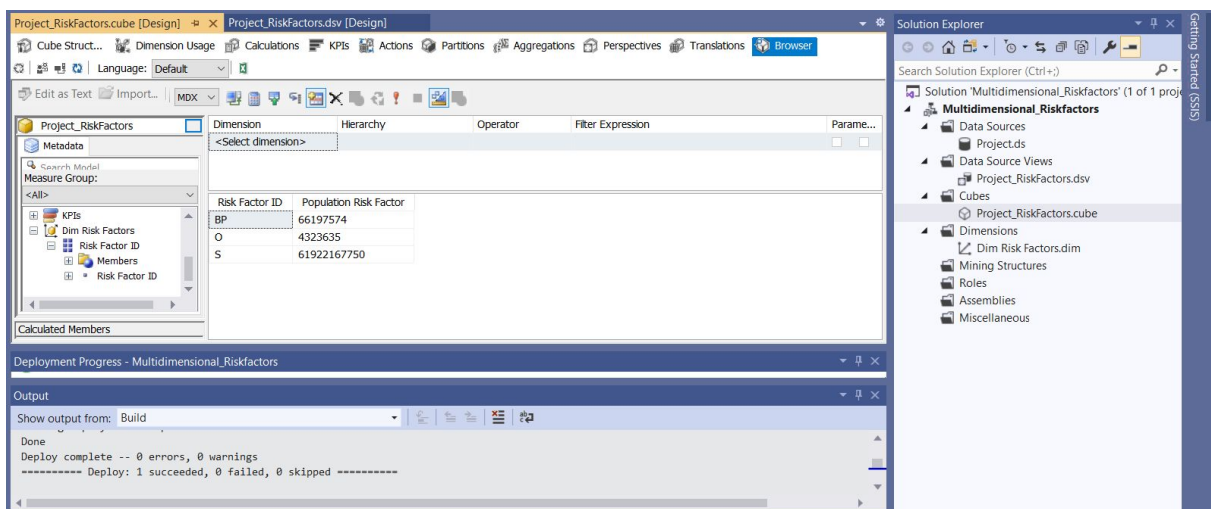
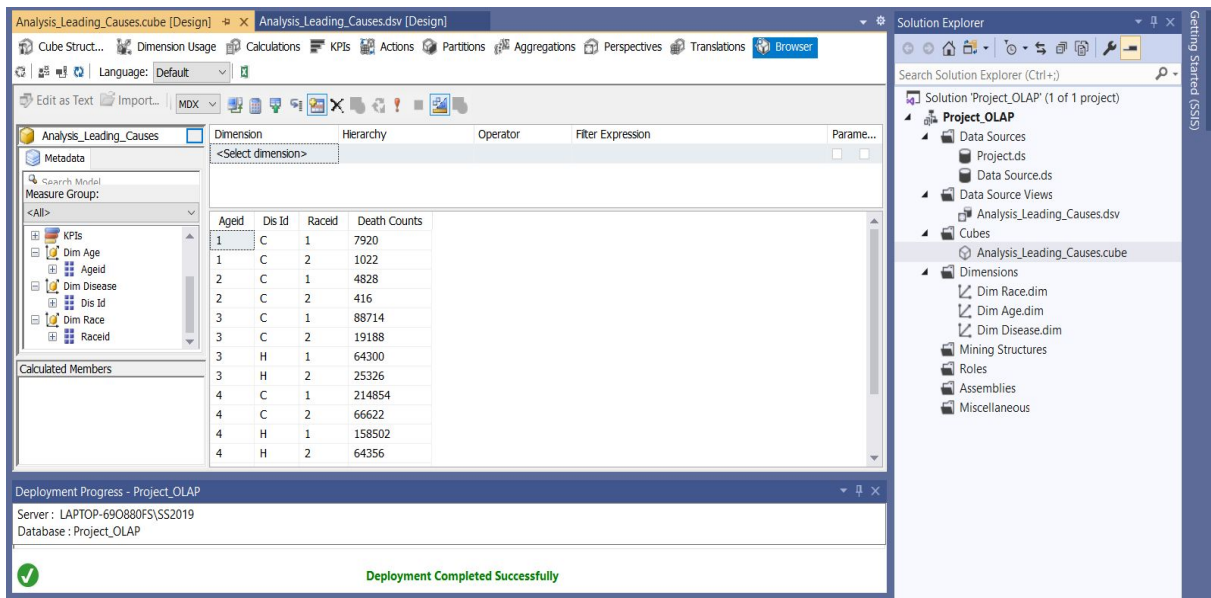
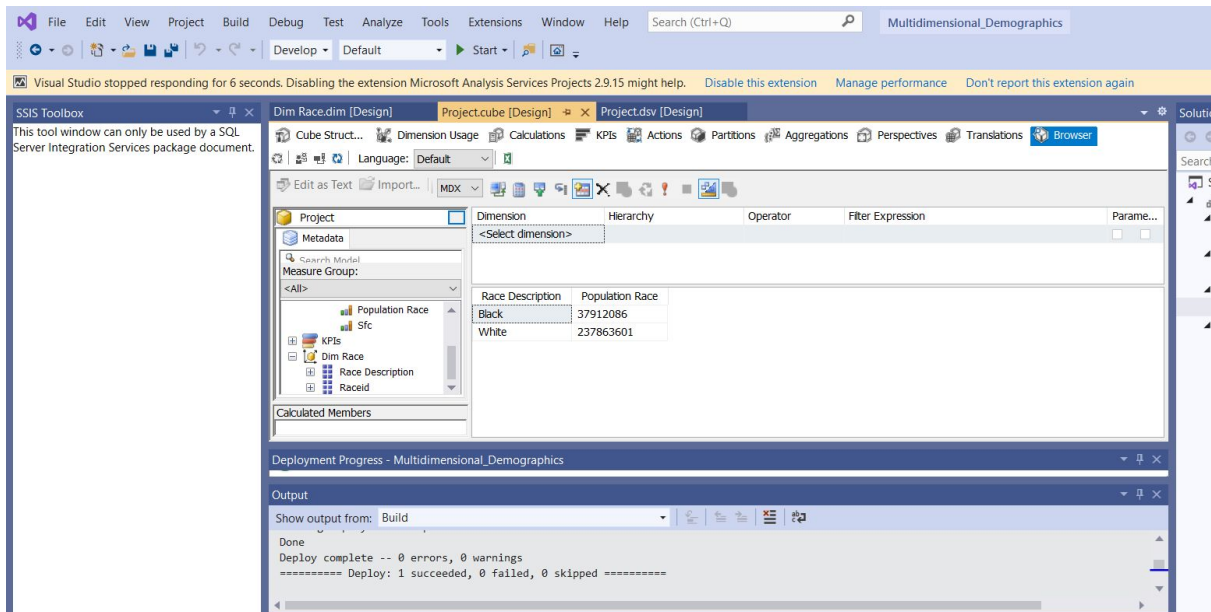


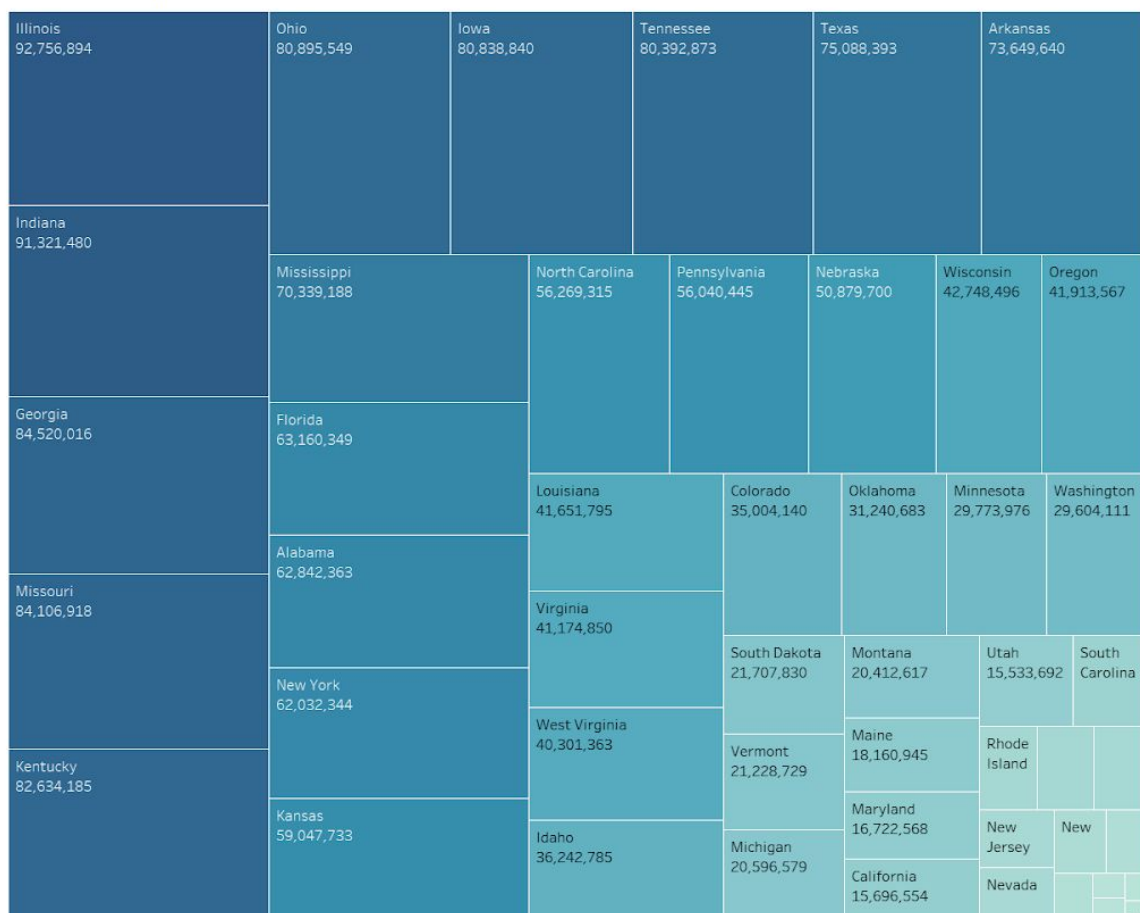
Fig 6: OLAP CUBES:

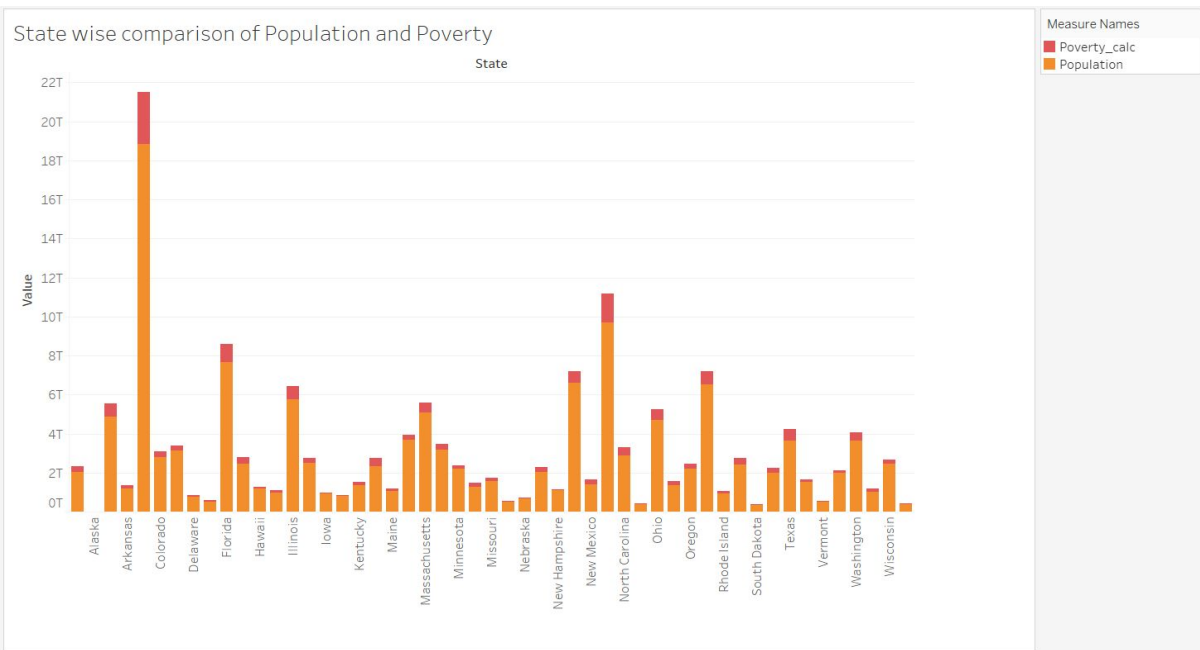
3.1 Reporting and Visualization Tools

Tableau is an interactive data visualization tool. It's an application that is used for translating raw data into valuable insights. A few advantages of using tableau over using any other visualization tool are :

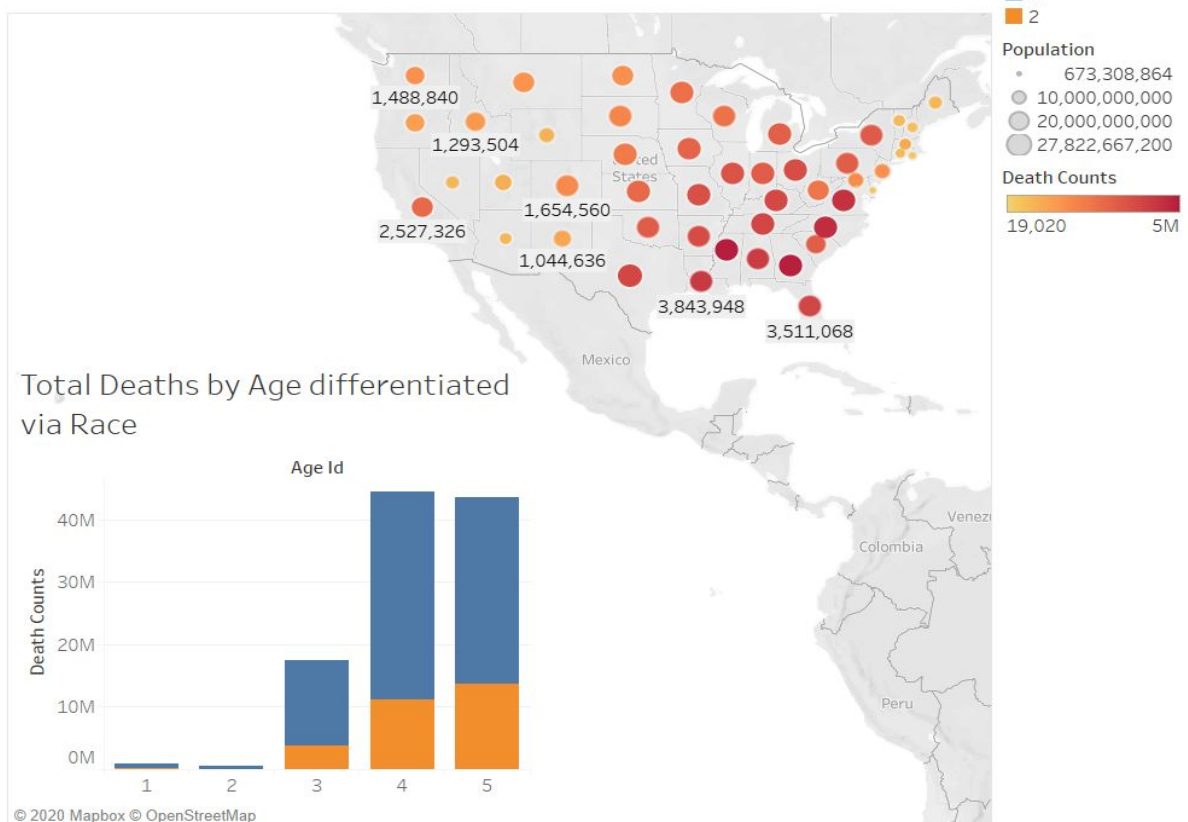
- Remarkable Visualization Capabilities
- Multiple Data Source Connections
- Ease of Use
- High Performance
- Mobile Friendliness

Uninsured Population Statewise

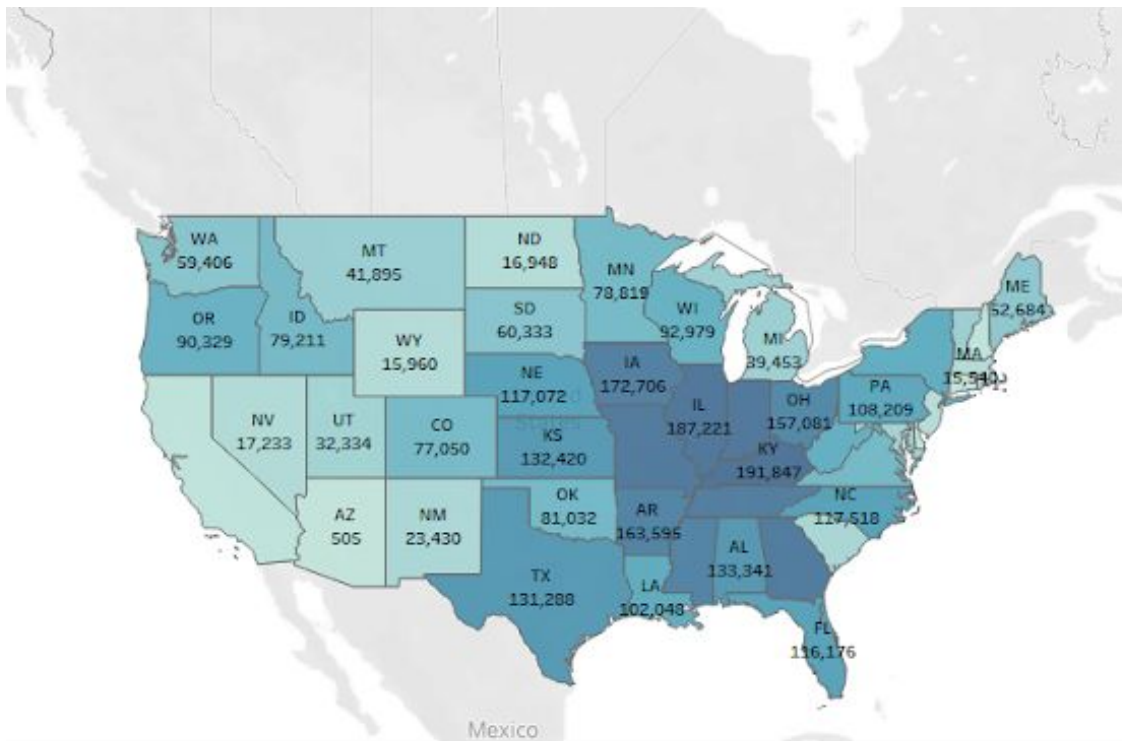
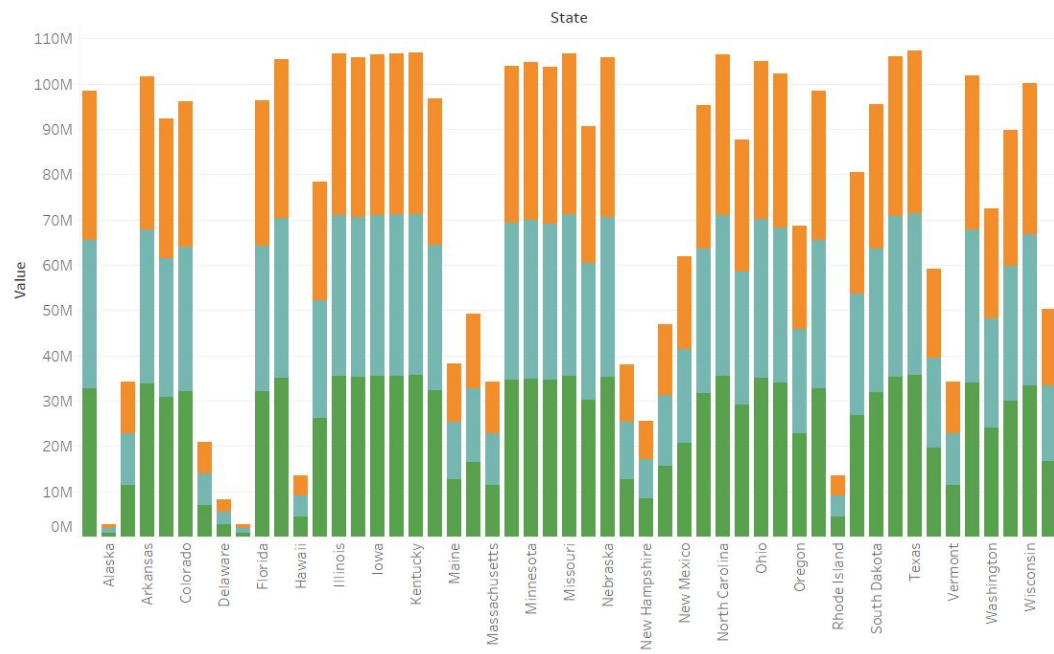




Total Deaths



Comparison of Population, Deaths, and Population with access to Healthcare



4.1 Interesting facts

This dataset provides key health indicators for local communities and encourages dialogue about actions that can be taken to improve community health (e.g., obesity, heart disease, cancer). The CHSI report contains over 200 measures for each of the 3,141 United States counties. The dataset covers factors like obesity, tobacco use, diet, physical activity, alcohol and drug use, sexual behavior and others substantially contribute to these deaths.

The dataset consists of detailed information regarding patients with diabetes, blood pressure, no. of smokers in each county, accessibility to medicaid centers, etc. These parameters form a platform for detailed and meaningful analysis.

The dataset contains figures regarding populations size, minimum and maximum population density, poverty values. This measure helps us to analyse the relation between the deaths, available services and other demographics.

We believed this dataset being detailed and covering multiple aspects helps us effectively analyse the actions to improve community health.

Questions our analysis can address:

The dataset revolves around the health demographic at the county level of the United States. There are a number of insights that can be derived from the sources listed.

1. What diseases ail the age groups provided in the dataset?
2. At what amount are preventive measures being taken towards the various health risks the country is facing?
3. Geo-spatial distribution of various risk factors and access to care to quickly realize which areas are lagging behind.
4. The Error table could also tell us about data consistency coming from different counties. Suggesting which areas need to be more careful with data generation/collection.

Appendix

Screenshots from SSIS Implementation



Fig: Data from flat file(Leading_Causes) to staging table

| | STATE_FIPS_CODE | COUNTY_FIPS_CODE | STATE_NAME | COUNTY_NAME | STATE_ABBR | CANCER_WH_A1 | CANCER_BL_A1 | CANCER_OT_A1 |
|----|-----------------|------------------|------------|-----------------|------------|--------------|--------------|--------------|
| 10 | 5 | 119 | Arkansas | Pulaski | AR | 17 | 0 | 0 |
| 11 | 6 | 1 | California | Alameda | CA | 14 | 13 | 17 |
| 12 | 6 | 13 | California | Contra Costa | CA | 19 | 17 | 0 |
| 13 | 6 | 19 | California | Fresno | CA | 15 | 0 | 0 |
| 14 | 6 | 29 | California | Kern | CA | 13 | 0 | 0 |
| 15 | 6 | 37 | California | Los Angeles | CA | 20 | 10 | 23 |
| 16 | 6 | 47 | California | Merced | CA | 17 | 0 | 0 |
| 17 | 6 | 53 | California | Monterey | CA | 19 | 0 | 0 |
| 18 | 6 | 59 | California | Orange | CA | 23 | 0 | 21 |
| 19 | 6 | 61 | California | Placer | CA | 30 | 0 | 0 |
| 20 | 6 | 65 | California | Riverside | CA | 17 | 0 | 0 |
| 21 | 6 | 67 | California | Sacramento | CA | 13 | 0 | 13 |
| 22 | 6 | 71 | California | San Bernardino | CA | 15 | 0 | 0 |
| 23 | 6 | 73 | California | San Diego | CA | 27 | 0 | 22 |
| 24 | 6 | 77 | California | San Joaquin | CA | 15 | 0 | 0 |
| 25 | 6 | 79 | California | San Luis Obispo | CA | 20 | 0 | 0 |

Fig: Data in staging table(SQL DB)



Fig: staging → Data Conversion → Destination

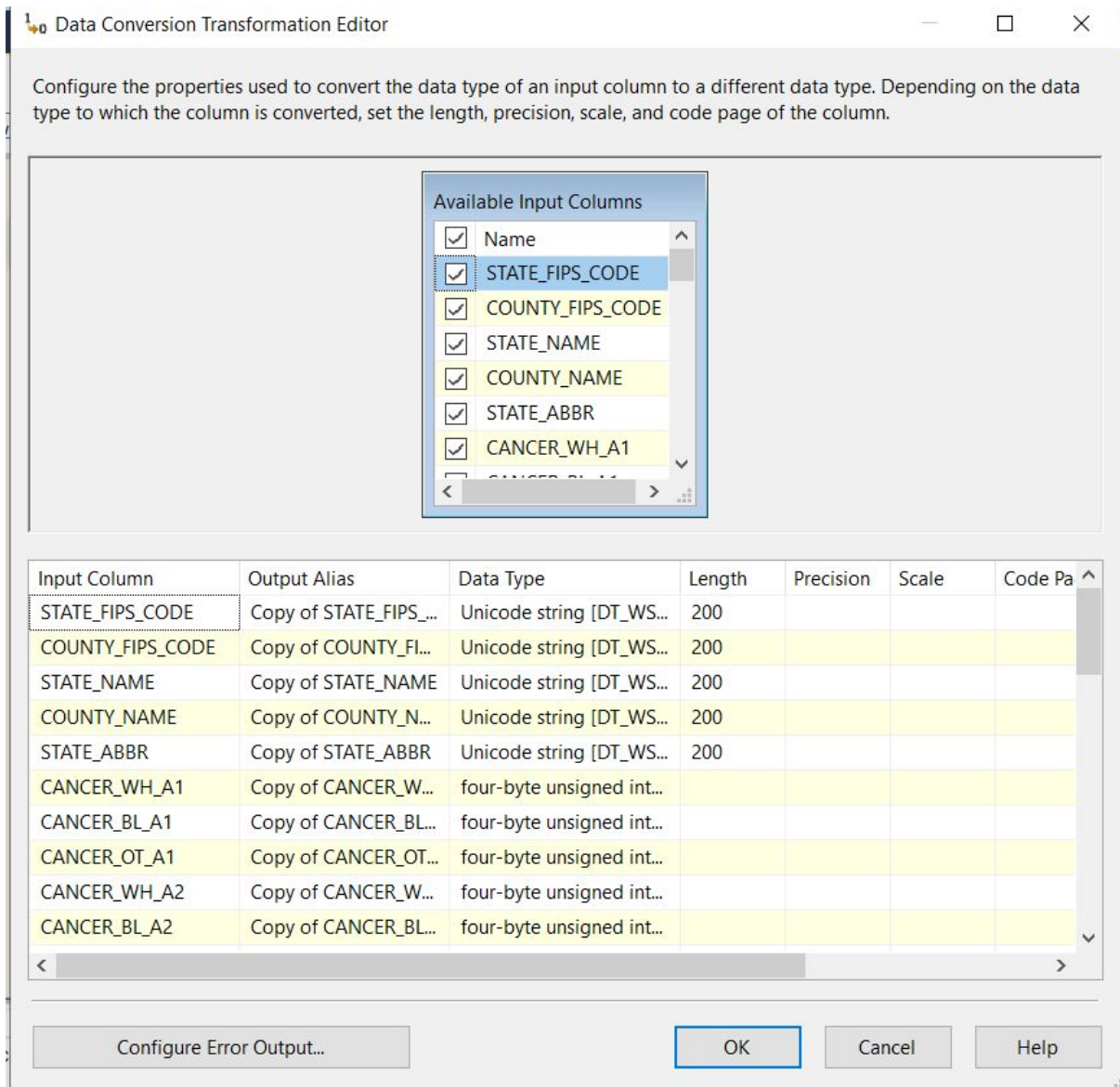


Fig: Data Conversion Transformation Editor



Fig: Dimension and Fact tables load - Leading causes of deaths

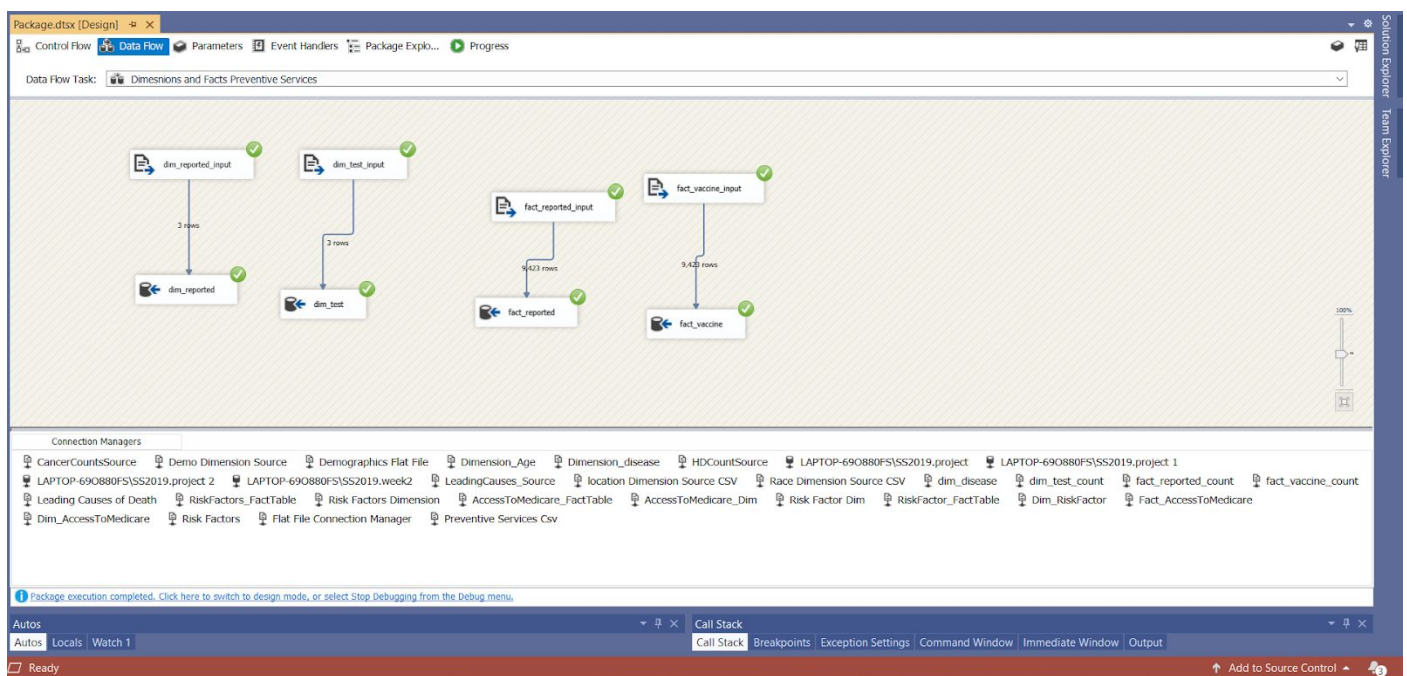


Fig: Fact and Dimension tables load for Preventive Measures.

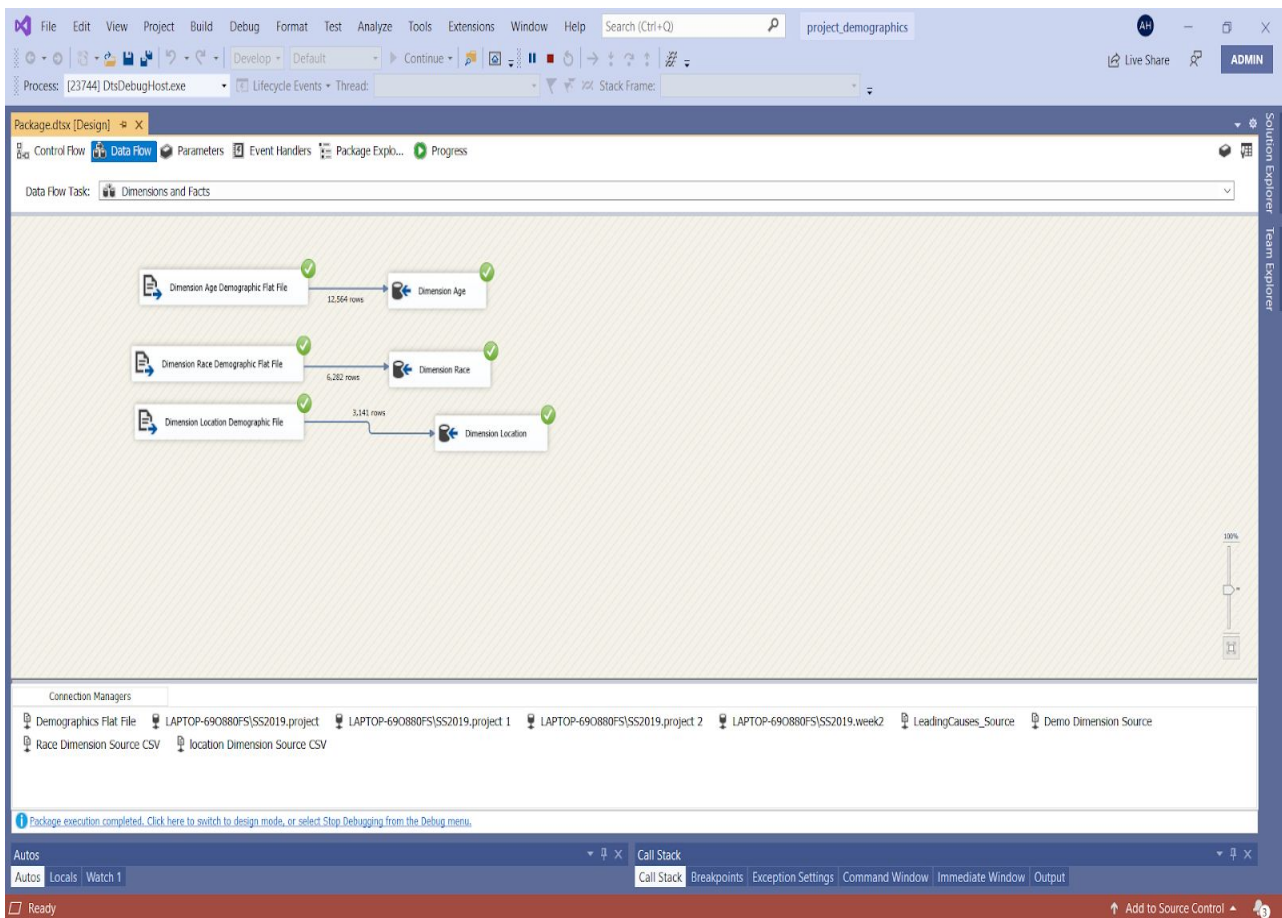


Fig: Fact and Dimension load- Demographics

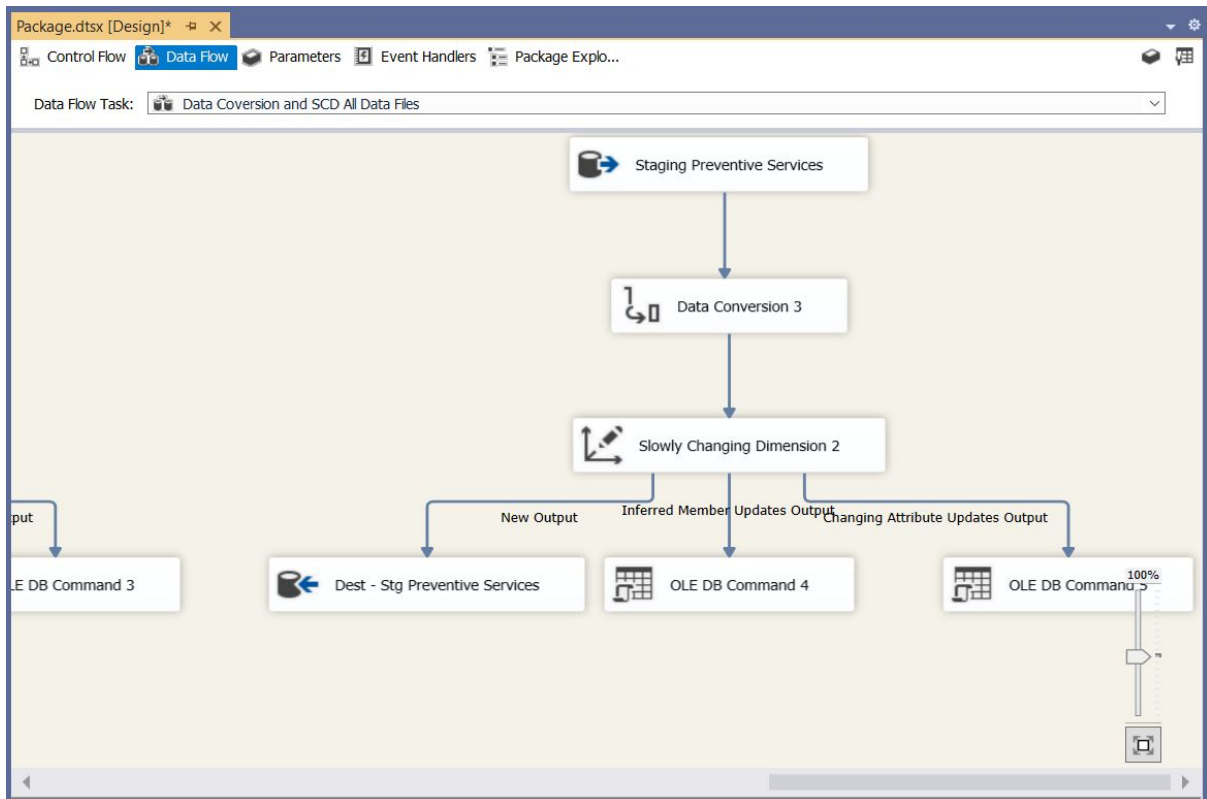


Fig: Slowly changing dimensions implemented for Preventive Services.

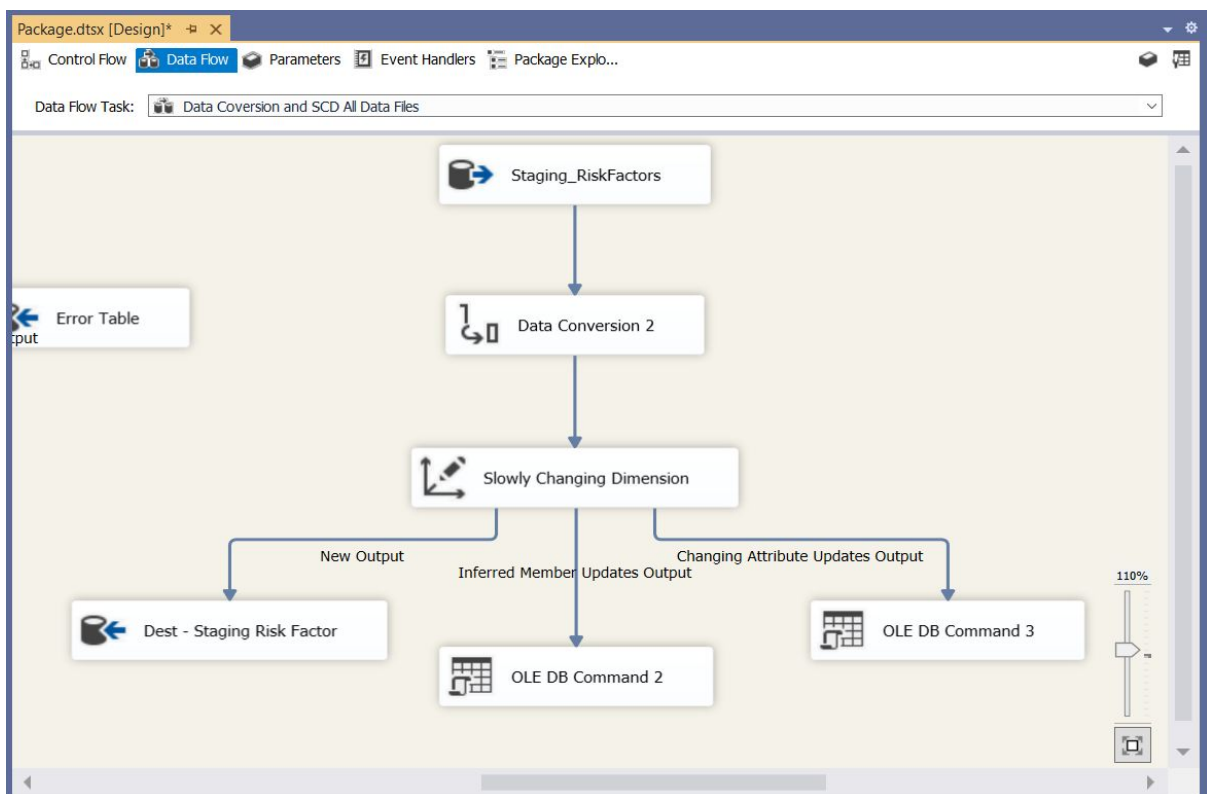


Fig: Slowly changing dimensions implemented for Risk Factors.

Professor's Comments:

| Sr No. | Date | Proposal | Comments |
|--------|------------|-------------------------|---|
| 1 | 11/11/2020 | Project Proposal 1 | Looks good Don't understand the transformations section |
| 2 | 27/11/2020 | Project Proposal 2 | <p>I wanted my comments in the end of the document as an appendix</p> <p>I don't expect to see my prompts in your document. You should use your own titles for the sections etc.</p> <p>Typically the cube will be built of the mart</p> <p>Race / Age are better as dimensions</p> <p>What is field / type</p> |
| 3 | 12/06/2020 | Project Design Document | <p>I still see my prompts as your titles</p> <p>You list dimension but sum of them will combine into a single dimensions age for example will not be 3 dimensions</p> <p>State code is a char(2) not a varchar</p> <p>In your final design it</p> |

| | | | |
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| | | | <p>would be best to have an indicator of race and the counts associated with them as opposed to individual fields with race as part of the name...</p> <p>Data model is not consistent with regards to case and naming</p> <p>I don't think it is fair to say ssis lacks that feature</p> <p>Validation isn't meant to be a manual process I'm not sure of your intentions based on your document</p> <p>You visualizations don't really tell me what is happening you need to make sure you have proper labels etc.</p> <p>I expect more visualizations</p> |
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