HOSPITAL CLAIM MANAGEMENT SYSTEM ANALYSIS

Created by Batch 75

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Business Challenge / Requirement

This is a Hospital Claim Management System which has many databases stored like Patient information, Claim information, Insurance Company Names, Hospital Details etc. So, to correctly manipulate and analyse this huge sets of data, the authority has requested for our help in Big Data Analytics, for handling and analysing of this datasets.

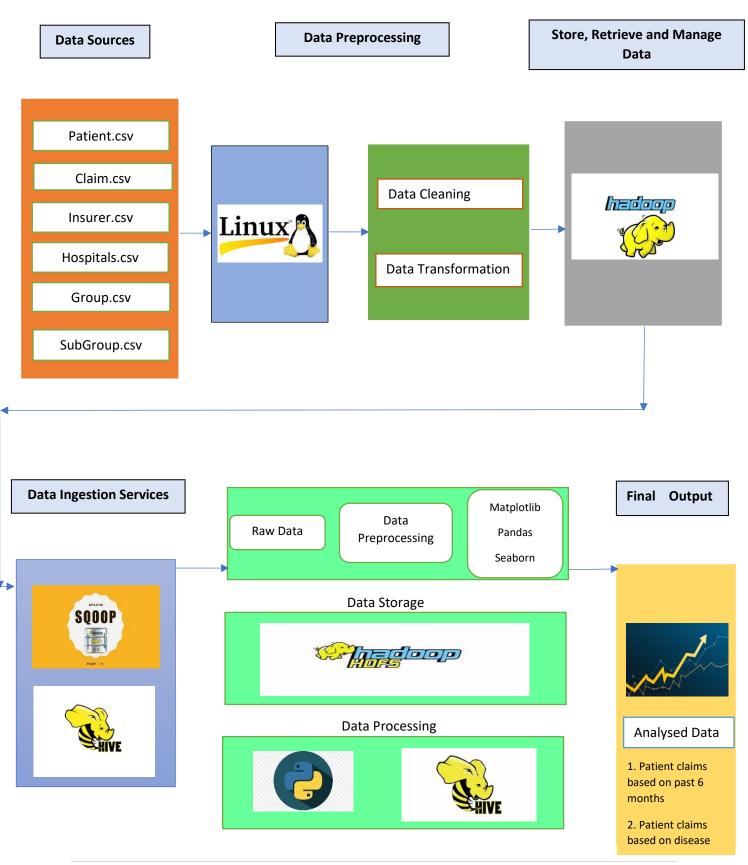
Goal of the Project

To meet our client requirements, we have to use many databases which store patient information, claim information, insurance company names etc. We need to do data cleansing to fix errors, duplicates and irrelevant data from the raw dataset. We have processed, filtered, transformed and analysed these data. Finally, we will plot a graph of 6months claim of patient and claims based on disease.

Project Architecture

- 1. A linux file server receives 14 files in total in Comma Seperated Value (CSV) format from the Hospital, Insurance Company and other sources. These are the source files that we will be using for our project.
- 2. Then we use data filtering in python to filter out NULL and invalid values from our data sources.
- 3. Now transformational logic is applied on the cleansed files and the data are imported to MYSQL for storage.
- 4. The file data and tables are validated, enriched, analysed, processed before loading into HDFS and HIVE.
- 5. Finally, after analysing of the tables and datasets, we are able to visualize the graph we need to get the relation between insurance claims and age group.

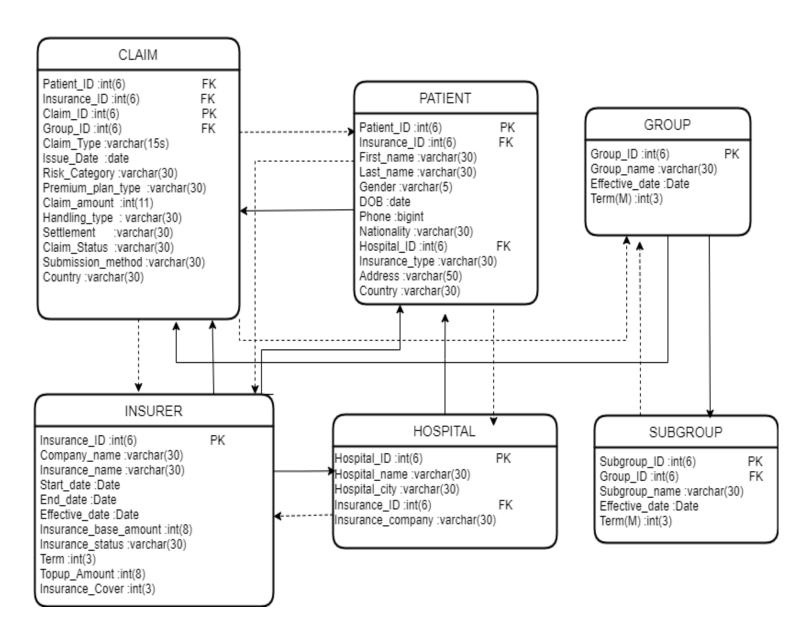
HOSPITAL CLAIM MANAGEMENT SYSTEM ANALYSIS ARCHITECTURE



❖ Datasets and Schemas

- 1. Data coming from third party sources reside in our local directory of Unix and in CSV format.
- 2. A final master table is created from existing tables which is then inserted in HDFS using SQOOP.
- 3. Lastly, that table is analysed and visualized using python data analytical package, and matplotlib to map the datasets.

❖ DATA MODEL



Description of the data source files

Claim CSV file fields

• Claim_ID : Unique, not null Primary Key

Patient_ID: Unique, not null Primary key for Patient table

• Insurance_ID : Unique, not null Primary Key for Insurer table

Group_ID: Unique, not null Primary Key for Group table

Claim_type : Shows the type of claim

• Issue date : Issue date for the claim made by patient

Risk_Category : Level of risk

• Premium_plan_type : Shows the frequency of subscription

• Claim_Amount : Amount of claim made

Handling_type : Type of handling

Settlement : Status of settlement of claimClaim status : Status of the claim made

• Submission_method : Method of claim submission

• Country : Country

Patient CSV file fields

• Patient ID : Unique, not null Primary Key

Insurance_ID: Unique, not null Primary Key for Insurer table
Hospital_ID: Unique, not null Primary Key for Hospital table

First_name : First name of the patientLast_name : Last name of the patient

• Gender : Gender

DOB : Date of birth of patientPhone : Contact number of patient

• Nationality : Nationality of patient

• Insurance_type: Term of the insurance covered

Address : Full address of patient

Country : Country in which the patient belong to

4 Insurer CSV file fields

Insurance_ID : Unique, not null Primary Key for Insurer table

• Company_name : Name of Company providing insurance

• Insurance_name : Name of Insurance Plan

Start_Date : Date of start of insurance plan in dd/mm/yyyy

End_Date : End Date of the plan in dd/mm/yyyy

• Effective_Date : Effective Date of the plan in dd/mm/yyyy

• Insurance_Base_Amount: Basic Coverage amount of the plan

• Term : Validity time period of plan in Years

• Topup_Amount : Amount of topup that can be added

Insurance_Cover : Head coverage under the plan

Hospital CSV file fields

Hospital_ID : unique, not null Primary key

Hospital Name : Name of the Hospital

• Hospital_City : City where the hospital is located

• Insurance_ID : Unique ID of Insurance as foreign key

• Insurance_Company : Name of Insurance Company

4 Group CSV file fields

Group_ID : Unique, not null Primary key of group table

• Group_Name : Name of the group of Disease

• Effective Date : Effective date

• Term : Term of validity in Months

SubGroup CSV file fields

SubGroup_ID : Unique, not null primary key

Group_ID : Unique , foreign key of table Group

• SubGroup_Name: Name of the disease under the parent disease

• Effective_Date : Effective date

• Term : Term of validity in months

❖ Description of the final file

Final CSV file fields

Patient_name : Name of the Patients

• Patient_address : Address of the Patients

• Claim_id : Unique, not null primary key

Insurance_details : Name of the insurance and insurance company

• Claim_status_details : Status of the claim made

• Claim_issue_date : Issue date for the claim made by patient

• Insurance_term : Validity time period of plan in Years

Patient_age : Age of Patient

• Group_details : Name of the disease and group of disease

• Insurance_amount : Base amount and top up amount

❖ Problem Statement

From this huge dataset, we process and analyze the data carefully and find out the following pattern:

- I. Insurance claims made by patient who are admitted in the Hospital from the past 6 months by the patient's age.
- II. Claims of patient based on disease.
- III. Claim uses and balance amount based on Insurance company.

Code Templates

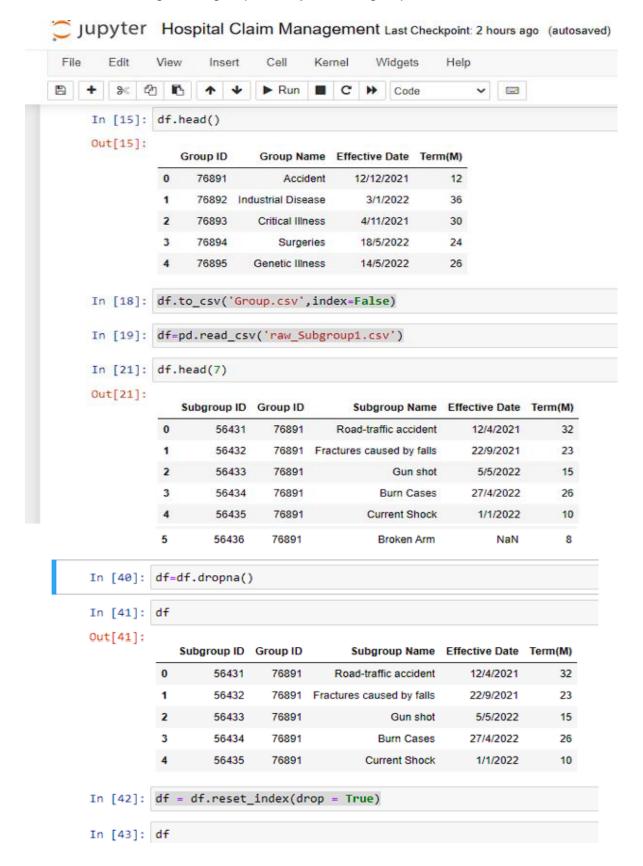
Data Processing:

First we fetch all the raw source files and then cleanse it using python pandas, such as removing duplicate values, null values, then it gets converted to processed data. Here is some code snippets.

Cleansing of group.csv

```
In [1]:
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
In [12]: df=pd.read_csv('raw_Group.csv')
In [13]: df.head()
Out[13]:
              Group ID
                           Group Name Effective Date Term(M) Address
                 76891
                               Accident
                                           12/12/2021
                                                           12
                                                                  NaN
                 76892 Industrial Disease
                                             3/1/2022
                                                          36
                                                                  NaN
           1
                                            4/11/2021
                 76893
                           Critical Illness
                                                          30
                                                                  NaN
                 76894
                                            18/5/2022
                              Surgeries
                                                          24
                                                                  NaN
                 76895
                          Genetic Illness
                                            14/5/2022
                                                          26
                                                                  NaN
In [14]: del df['Address']
In [15]:
          df.head()
Out[15]:
              Group ID
                           Group Name Effective Date Term(M)
```

Cleansing of subgroup files by removing duplicates.



Cleansing of subgroup files by removing null values.

```
In [66]: df=pd.read_csv('raw_Subgroup5.csv')
In [67]: df
Out[67]:
               Subgroup ID
                           Goup ID
                                      Subgroup Name Effective Date Term(M)
           0
                    56470
                            768925
                                              Autism
                                                              NaN
                                                                        25
                    56471
                            768925
                                      Down syndrome
                                                         23/1/2022
            1
                                                                        30
            2
                    56472
                            768925 FragileX syndrome
                                                         22/9/2021
                                                                        27
            3
                    56473
                            768925
                                      Turner syndrome
                                                          4/4/2022
                                                                        20
                    56474
                            768925
                                          Trisomy 18
                                                         27/4/2022
                                                                        28
            5
                    56475
                            768925
                                           Trisomy 13
                                                          1/1/2022
                                                                        14
In [68]: df=df.dropna()
In [69]: df
Out[69]:
               Subgroup ID Goup ID
                                      Subgroup Name Effective Date Term(M)
           1
                    56471
                            768925
                                      Down syndrome
                                                         23/1/2022
                                                                        30
            2
                    56472
                            768925 FragileX syndrome
                                                         22/9/2021
                                                                        27
```

Processed data

	Patient ID	Insurance ID	First name	Last name	Gender
0	45098.0	1003.0	Akriti	Singh	F
1	45099.0	1003.0	Pritam	Bannerjee	M
2	45100.0	1003.0	Salim	Khan	М
3	45101.0	1003.0	Yasoof	Akhtar	M
4	45102.0	1003.0	Wahida	Begum	F
5	45103.0	1003.0	Saheb	Chowdhury	М
6	45104.0	1003.0	Liza	Kusari	F
7	45105.0	1003.0	Rahul	Parmar	М
8	45106.0	1003.0	Koyel	Singh	F
9	45107.0	1003.0	Rai	Sen	F
10	45108.0	1003.0	Saina	Das	F
11	45109.0	1003.0	Saili	Goyel	F
12	45110.0	1003.0	Lianna	Reddy	F
13	45111.0	1003.0	Sanvi	Shetty	F
14	45112.0	1003.0	Sourav	Sanyal	М
15	45113.0	1003.0	Rounak	Bansal	М
16	45114.0	1003.0	Diya	Gupta	F
					12

12 | Page

	Phone no	Nationality	Hospital ID	Insurance Type
0	7.405288e+09	American	70001.0	Long Term
1	4.014207e+09	American	70005.0	Short Term
2	3.206315e+09	American	70004.0	Long Term
3	4.057383e+09	American	70002.0	Long Term
4	7.852044e+09	American	70003.0	Short Term
5	8.567862e+09	American	70005.0	Short Term
6	2.033718e+09	American	70004.0	Long Term

alhost:8888/nbconvert/html/Untitled15.ipynb?download=false

8/22,	10:25 AM			Untitled15
7	6.176708e+09	American	70003.0	Short Term
8	7.087782e+09	American	70002.0	Short Term
9	6.204418e+09	American	70001.0	Short Term
10	6.263711e+09	American	70001.0	Long Term
11	3.472706e+09	American	70003.0	Short Term
12	6.172935e+09	American	70005.0	Short Term
13	9.319498e+09	American	70002.0	Long Term
14	2.679530e+09	American	70004.0	Short Term
15	8.136449e+09	American	70004.0	Long Term
16	6.192009e+09	American	70002.0	Short Term
17	4.804396e+09	American	70003.0	Long Term
18	3.014947e+09	American	70001.0	Long Term
19	7.142611e+09	American	70005.0	Short Term
20	8.125245e+09	American	70004.0	Short Term

	Address	Country
0	1660 Meadow DriveTennessee 27	USA
1	4427 Green hill road, Arkanas 08	USA
2	776 Harper street, kentucky 65	USA
3	53 Flynn Street Nebraska 16	USA
4	3085 simons road pennysylvania 34	USA
5	2929 forest drive kansas 05	USA
6	3 Herald New York 10001-3065 USA	USA
7	42 W 35TH New York 10001-2233	USA
8	213 W 35th New York NY	USA
9	350 W 31st New York	USA
10	939 Stadium Drive Massachusetts	USA
11	3319 HORSESHOE LANE CALIFORNIA	USA
12	4170 POPLAR STREET, ILLIOIS	USA
13	1135 NORMAN STREET, CALIFORNIA	USA
14	3432 SYCAMORE STREET, CALIFORNIA	USA

Insurer.csv after processing and cleansing

	Insurance ID	Company Name	Insurance Name	Start Date(dd/mm/yyyy)	End Date(dd/mm/yyyy)	Effective date(dd/mm/yyyy)	Insurance Base Amount	Insurance Status	Term(Yrs)	TopUp Amount	
1	1001	Aegon Life Insurance Co. Ltd.	Individual Health Insurance	1/1/2020	1/1/2036	15/1/2020	200000.0	Active	15	200000	
4	1002	Edelweiss Life Insurance Co.	Family Health Insurance	3/1/2020	3/1/2039	15/3/2020	300000.0	Active	18	300000	
7	1003	Birla Sun Life Insurance Co.	Critical illness Insurance	5/2/2020	5/2/2041	17/05/2020	400000.0	Active	20	150000	

Importing Processed data to MYSQL

List of all tables, that are imported to MySQL from the processed files.

```
Database changed
mysql> show tables;
+-----+
| Tables_in_project |
+-----+
| Claim |
| Hospital |
| Insurer |
| Patient |
| groups |
| subgroup |
+-----+
6 rows in set (0.00 sec)
```

• CLAIM TABLE

Field	Туре	Null	Key	Default	Extra
Patient ID	int(6)	YES	MUL	NULL	
Insurance ID	int(6)	YES	MUL	NULL	
Claim ID	int(6)	NO	PRI	NULL	
Group_ID	int(6)	YES	MUL	NULL	
Claim_type	varchar(15)	YES	i i	NULL	
Issue_date	date	YES	i i	NULL	
Risk_Category	varchar(30)	YES	İ	NULL	
Premium_plan_type	varchar(30)	YES	İ	NULL	
Claim_amount	int(11)	YES	İ	NULL	
Handling_type	varchar(30)	YES		NULL	
Settlement	varchar(30)	YES		NULL	
Claim_status	varchar(30)	YES		NULL	
Submission_method	varchar(30)	YES		NULL	
Country	varchar(30)	YES		NULL	

	tient_ID Insu e Settlement	Claim_statu	s Submiss	ion_meth	od Country	•		Risk_Category +					Handling
					++ Cashless	· 2021-1			Monthl		i		Complex
ı i	In Process		Interne		India		'			,			
	45106	1003	4009	76894	Reimbursement	2022-0	5-09	High	Semi-a	nnualy	1	29000	Acceler
Ιd	Invalid	INVALID	Mail		India								
	45111	1003	4014	76893	Cashless	2022-0	1-03	Low	Monthl	y	1	35000	Standar
L	Settled	RECOVERED	Interne	t	India								
	45113	1003	4016	76895	Reimbursement	2021-1	1-12	High	Monthl	y	1	20500	Standar
L	Settled	ACTIVE	Interne	t	India								
	45117	1003	4020	76891	Reimbursement	2021-1	1-04	Low	Quater	ly	1	24100	Standar
L	Settled	RECOVERED	Mail		India								
	45121	1003	4024	76892	Cashless	2022-0	4-30	Low	Quater	ly	1	19999	Express
L	In Process		Mail		India								
	45123	1003	4026	76891		2022-0	5-18	High	Semi-a	nnualy	1	49500	Acceler
Ιd	In Process		Call		India								
	45127	1003	4030	76894	Cashless	2022-0	4-21	High	Semi-a	nnualy	1	41000	Standar
l .	Invalid	INVALID	Mail		India								
	45128	1003	4031	76892	Reimbursement	2022-0	5-14	High	Quater	Ly	1	32000	Acceler
Iq	In Process		Call	7.004	India							50000	
	45134	1003	4037	76891	Reimbursement	2022-0	4-07	Low	Semi-a	nnualy	1	59000	Standar
1		RECOVERED	Mail	76002 1	India	1 2022 0	4 04	I Manuald com	I Manabal			20000	Louis
	45135 Invalid	1003 INVALID	4038 Mail	70892	Reimbursement India	2022-0	4-01	Medium	Monthl	y	1	20000	Express
١	45139	1003	4042	76001 I	Cashless	1 0000-6		Llow	Semi-a	anualy.	4	45000	Standar
	In Process		Mail	70091	India	0000-0	0-00	LOW	l seur-a	iliuaty	1	43000	Scaling
١	45142	1003	4045	76895 I	Cashless	2022-6	3-31	l Medium	Quater	lv	100	50000	Express
ı i		INVALID	Interne		India	1 2022-0	J-J1	i ilea cari	Quacer	Ly	1	30000	Lypicss
'	45145 I	1003 I	4048 I		Cashless	1 2022-0	5-08	Low	Monthl	v	1	37000	I Standar
Ι	Settled	RECOVERED	Interne		India	1 2022 0			1				1 30311331
	45147 I	1003	4050 I		Reimbursement	1 2022-6	4-23	Low	Quater	lv	1	22500	Standar
ī	In Process		Mail		India	,				-,			,

• INSURER TABLE

```
mysql> describe Insurer:
                  | Type | Null | Key | Default | Extra |
 Field
                   Insurance ID
                   | varchar(30) |
| Company_name
                                 YES
                                            NULL
| Insurance_name
                   | varchar(30) |
                                 YES
                                           NULL
                                 YES |
 Start_date
                    | date
                                            NULL
                    | date
                                 YES |
 End_date
                                          NULL
                    date
                                 YES |
| Effective date
                                          NULL
                                YES I
 Insurance_base_amount | int(8)
                                          NULL
                                          NULL
 Insurance status | varchar(30) | YES |
                                 YES
                                          NULL
                    | int(3)
 Term
 Topup_Amount
                               YES
                                          NULL
                    | int(8)
                                | YES |
 Insurance cover
                    | int(4)
                                           NULL
11 rows in set (0.00 sec)
```

• PATIENT TABLE

Field	Type	Null	Key	Default	Extra
Patient_ID	int(6)	NO NO	PRI	NULL	
Insurance_ID	int(6)	YES	MUL	NULL	į
First_name	varchar(30)	YES	i i	NULL	İ
Last_name	varchar(30)	YES	i i	NULL	
Gender	varchar(5)	YES	j i	NULL	
DOB	date	YES	l i	NULL	
Phone	bigint(20)	YES	i i	NULL	
Nationality	varchar(30)	YES	i i	NULL	
Hospital_ID	int(6)	YES	MUL	NULL	
Insurance_type	varchar(30)	YES	<u> </u>	NULL	l
Address	varchar(50)	YES	l i	NULL	
Country	varchar(30)	YES	l i	NULL	

Patient_ID Insurance_ID ess	Country	i -		DOB						Insurance_type	
	+		-+	+		-+-		+			-+-
		Singh	F	2002-05-09	8908908398	Τ	Indian	I .	90001	Longterm	1
sp road kolkata 64	India										
	•	Bannerjee	M	2001-02-02	9087988092	1	Indian	1	90003	Short term	- 1
p Mukherjee road kolkata											
45100 1003	•	Khan	M	2001-06-09	9900887766	Т	Indian	Γ	90005	Long term	-1
oy road howrah 34											
	•	Akhtar	M	2000-08-07	9988455767	Т	Indian	I	90002	Long term	- 1
	India		1.6	1 2002 07 00	0000433467		T-44		00003	Chart torn	ī
45102 1003 roy road karnataka 04	Wahida India	Begum	F	2003-07-08	9090123407	1	Indian	ı	90003	Short term	, ,
		Chowdhury	I M	1992-03-09	0207200070	4	Indi an		90004	Short term	1
urnadas road kolkata 67	India	Cilowaliai y	1 "	1992-03-09	3601630313	1	Tilocali	1	30004	SHOLL CELL	'
		Kusari] F	1990-04-10	9089757868	1	Indian	1	90004	Long term	-1
	India	1 110501 0		1 2330 0. 20					,,,,,	Long com	
	Rahul	Parmar	I M	2000-09-06	9999977685	1	Indian	I .	90001	Short term	-1
inson road kolkata 98	India										
45106 1003	Koyel	Singh	F	2001-09-09	9999963667	1	Indian	1	90005	Long term	-1
oy road kolkata 98	India										
	•	Sen	F	2002-09-08	9864680066	1	Indian	1	90002	Short term	-1
	India										
	•	Das	F	1999-02-09	9876543210	Т	Indian	I	90003	Long term	-1
oy road kolkata 89	India	I Carral	1.5	I 4007 03 00	0630537440		T		00004	Chart too	
	•	Goyel	F	1997-03-08	9638527410	Т	Indian		90001	Short term	-1
s road Bombay 01 45110 1003	India Lianna	Reddy	I F	1992-07-22	05197/2360		Indian		98882	Long term	i
c road Malda 09	Llanna India	Reddy	1	1992-07-22	9516742500	1	Inditall		30002	Long term	
45111 1003		Shetty	I F	1995-03-22	9638521470	1	Indian		90003	Long term	- 1
road punjab 99	India	Jilecey		1555-05-22	7030321470	٦,			70003	Long CCTH	
		Sanyal	I M	1980-09-13	9032165478	Ī.	Indian		90005	Long term	1
uresh roy road kolkata 9											

• GROUPS TABLE

Field	Type				
Group_ID Group_Name Effective_date	int(6) varchar(30)	1	NO YES	PRI 	

	Group_ID	· —	Effective_date	Term
		Accident	2021-12-12	12
	76892	Industrial Disease	2022-03-01	36
	76893	Critical Illness	2021-04-11	30
	76894	Surgeries	2022-05-18	24
ı	76895	Genetic Illness	2022-05-14	26

HOSPITAL TABLE

```
mysql> describe Hospital;
 Field
                                   | Null | Key | Default | Extra
                      Type
 Hospital ID
                     | int(6)
                                   NO
                                            PRI | NULL
                     | varchar(30)
 Hospital_name
                                     YES
                                                  NULL
 Hospital City
                     | varchar(30)
                                     YES
 Insurance ID
                     | int(6)
                                     YES
                                            MUL
                                                  NULL
 Insurance Company | varchar(30) | YES
                                                  NULL
 rows in set (0.00 sec)
```

```
mysql> create table Hospital( Hospital_ID int(6) Primary Key, Hospital_name varchar(30), Hospital_City varchar(30), Insurance_ID in
t(6), Insurance_Company varchar(30) );
Query OK, 0 rows affected (0.04 sec)
nysql> load data local infile "Desktop/Project/Hospitals.csy" into table Hospital fields terminated by "," lines terminated by "\n"
 ignore 1 rows:
Query OK, 15 rows affected, 2 warnings (0.01 sec)
Records: 15 Deleted: 0 Skipped: 0 Warnings: 2
mysql> select * from Hospital;
                                                                          | Insurance_ID | Insurance_Company
| Hospital_ID | Hospital_name
                                                     | Hospital_City
         70001
                 The Johns Hopkins Hospital
                                                       Baltimore MD
                                                                                     1001
                                                                                             Aegon Life Insurance Co. Ltd.
         70002
                  Massachusetts General Hospital
                                                       Boston MA
                                                                                      1001
                                                                                             Aegon Life Insurance Co. Ltd.
         70003
                  UCSF Medical Center
                                                       San Francisco CA
                                                                                     1001
                                                                                             Aegon Life Insurance Co. Ltd.
         70004
                  Mayo Clinic
                                                       Phoenix AZ
                                                                                      1001
                                                                                             Aegon Life Insurance Co. Ltd.
                 Brigham And Women's Hospital
Rechts der Isar Hospital
         70005
                                                       Boston MA
                                                                                     1001
                                                                                             Aegon Life Insurance Co. Ltd.
                                                                                             Edelweiss Life Insurance Co.
         80001
                                                       Munich
                                                                                     1002
         80002
                  University Hospital Cologne
                                                       Cologne
                                                                                     1002
                                                                                             Edelweiss
                                                                                                         Life Insurance Co.
                  University Medical Center
                                                       Freiburg
         80003
                                                                                     1002
                                                                                             Edelweiss
                                                                                                         Life Insurance Co.
                  University Medical Center Schl
                                                       Keil
                                                                                     1002
                                                                                             Edelweiss
                                                                                                         Life Insurance Co.
         80004
                  University Hospital Regensburg
                                                       Regensburg
                                                                                             Edelweiss
         80005
                                                                                     1002
                                                                                                         Life Insurance Co.
                 Rainbow Hospitals
Apollo Hospital
                                                                                             Birla Sun Life Insurance Co.
Birla Sun Life Insurance Co.
         90001
                                                       Hyderabad
                                                                                     1003
         90002
                                                       Visakhapatnam
                                                                                     1003
                  Gauhati Medical College and Ho
                                                       Gauhati
                                                                                             Birla Sun Life Insurance Co.
         90003
                                                                                     1003
         90004
                  ATIMS
                                                       Patna
                                                                                     1003
                                                                                             Birla Sun Life Insurance Co.
         90005
                 Ruby General Hospital
                                                       Kolkata
                                                                                     1003
                                                                                             Birla Sun Life Insurance Co.
15 rows in set (0.00 sec)
```

• SUBGROUP TABLE

```
nysql> describe subgroup;

Field | Type | Null | Key | Default | Extra |

Subrgoup_ID | int(6) | NO | PRI | NULL | |

Group_ID | int(6) | YES | MUL | NULL | |

Subgroup_name | varchar(30) | YES | | NULL | |

Effective_date | date | YES | | NULL | |

Term | int(3) | YES | | NULL | |

Frows in set (0.00 sec)
```

ore 1 rows; ry OK, 25 ro	ows affected	d (0.01 sec)	up.csv" into tablo	e subgro	oup fields terminated by "," lines terminated by "
ords: 25 De	eleted: 0 S	Skipped: 0 Warnings: 0			
ql> select '	_				
Jbrgoup_ID		+ Subgroup_name	+ Effective_date +	Term	
56431	76891	Road-traffic accident	2021-12-04	32	
56432	76891	Fractures caused by falls	2021-09-22	23	
56433	76891	Gun shot	2022-05-05	15	
56434	76891	Burn Cases	2022-04-27	26	
56435	76891	Current Shock	2022-01-01	10	
56441	76892	Industrial deafness	2022-01-23	32	
56442	76892	Respiratory Problems	2021-09-22	23	
56443	76892	Breathing Problems	2022-05-05	15	
56444	76892	Dermatitis	2022-04-27	26	
56445	76892	Musculoskeletal disorder	2022-01-01	10	
56451	76893	Cancer	2022-01-23	30	
56452	76893	Major organ transplant	2021-09-22	27	
56453	76893	Kidney failure	2022-05-05	15	
56454	76893	Paralysis	2022-04-27	26	
56455	76893	First heart attack	2022-01-01	10 i	
56461	76894	Neurological Surgery	2022-01-23	30	
56462	76894	Oncology	2021-09-22	27	
56463	76894	Chemotherapy	2022-04-04	20	
56464	76894	Hystoscopy	2022-04-27	15	
56465	76894	Nasal Concha	2022-01-01	10	
56471	76895	Down syndrome	2022-01-23	30	
56472	76895	FragileX syndrome	2021-09-22	27	
56473	76895	Turner syndrome	2022-04-04	20	
56474	76895	Trisomy 18	2022-04-27	28	
56475	76895	Trisomy 13	2022-01-01	14	

Final Table in MYSQL

```
nysql> select´* from Final_table limit 15;
| Claim_ID | Insurance_details
                                                                                                                                                    | Claim_stat
                                                                                                                      | Insurance amount |
| Liza Kusari | 34- ks sahani road. Mumbai 05 India
                                                                   | 4007 | Critical illness Insurance,Birla Sun Life Insurance Co. | In Process
                                                              32 | Genetic Illness,Down syndrome | 550000 |
|ia | 4060 | Critical illness Insurance,Birla Sun Life Insurance Co. | In Process
 ACTIVE | 2021-12-12 | 20 | 32
| Liza Kusari | 34- ks sahani road. Mumbai 05 India
 ACTIVE | 2022-01-14 | 20 |
Sanvi Shetty | 34-yk road punjab 99 India
                                                               32 | Accident,Road-traffic accident
                                                                                                                                    550000 |
| Sanvi Shetty | 2022-01-03 | 20 |
| Sanvi Shetty | 34-yk road punjab 99 India
| COVERED | 2022-03-03 | 20 |
| Khushi Khan | 9- ss roy road Delhi 09 India
| COVERED | 2021-11-04 | 20 |
                                                                     | 4014 | Critical illness Insurance,Birla Sun Life Insurance Co. | Settled,RE
                                                               27 | Critical Illness,Kidney failure | 550000 | 4069 | Critical illness Insurance,Birla Sun Life Insurance Co. | Settled,RE
                                                               27 | Industrial Disease,Dermatitis | 550000 |
| 4020 | Critical illness Insurance,Birla Sun Life Insurance Co. | Settled,RE
                                                               32 | Accident,Burn Cases
                                                                                                                                    550000 |
                                                               | 4073 | Critical lithess Insurance,

32 | Critical Illness,First heart attack | 550000 |

| 4026 | Critical illness Insurance,Birla Sun Life Insurance Co. | In Process

| 550000 |
 | Khushi Khan | 9-ss roy road Delhi 09 India
           2022-01-23
COVERED
                                              20 |
 | Mollika Singhania | 7- kk road Chennai 09 India
 ACTIVE
            2022-05-18
                                               20 |
| Mollika Singhania | 7- kk road Chennai 09 India
| ACTIVE | 2022-01-15 | 20 |
                                                                            4074 | Critical illness Insurance, Birla Sun Life Insurance Co. | In Process
                                                               20 | Surgeries, Nasal Concha
                                                                                                                                    550000 |
 Rohan Ghose | 9- linkon street kolkata 9 India
                                                                            4031 | Critical illness Insurance, Birla Sun Life Insurance Co. | In Process
          2022-05-14
                                                               29 | Industrial Disease, Musculoskeletal disorder | 550000 | 4090 | Critical illness Insurance, Birla Sun Life Insurance Co. | Invalid, IN
 ACTIVE
                                               20 |
 Rohan Ghose | 9- linkon street kolkata 9 India
                                                               29 | Accident, Fractures caused by falls
            2022-07-03
                                              20 |
                                                                                                                                    550000 |
VALID
 | Lokesh Ghose | 2- sr road Pune 4 India
                                                                            4038 | Critical illness Insurance, Birla Sun Life Insurance Co. | Invalid, IN
 .
/ALID
           2022-04-01
                                                               26 | Industrial Disease, Respiratory Problems
                                              20 |
                                                                                                                                    550000 |
4081 | Critical illness Insurance, Birla Sun Life Insurance Co. | In Process
                                                               26 | Surgeries,Oncology
                                         20 |
                                                                                                                                    550000 |
 Rani Chakraborty | 6 - Fort William street kolkata 9 India |
ACTIVE | 0000-00-00 | 20 | 30 | Acc
                                                                           4042 | Critical illness Insurance, Birla Sun Life Insurance Co. | In Process
                                             20 |
                                                               30 | Accident,Burn Cases
                                                                                                                                    550000 |
         Alam | 7- ju colony kolkata 7 India
| 2022-03-31 | 20 |
                                                                            4045 | Critical illness Insurance, Birla Sun Life Insurance Co. | Invalid, IN
 Afzal Alam
                                                               23 | Genetic Illness, Trisomy 13 | 550000 | 4048 | Critical illness Insurance, Birla Sun Life Insurance Co. | Settled, RE
VALID
 Shravan Ghosal | 7- cu road Chennai 8 India
                                                               27 | Surgeries, Neurological Surgery
 COVERED | 2022-05-08
                               | 20 |
                                                                                                                                    550000 I
 15 rows in set (0.00 sec)
```

DATA TRANSFORMATION

	ort DBA	SQLAIC	hemy connec ects are no	table(eng	ine/conne	ection) or	databas	se stri	ng URI or		ing: pandas BAPI2 conne	
[12]:	F	atient_ID	Insurance_ID	Hospital_ID	First_name	e Last_name	Gender	DOB	Phone	Nationality	Insurance_type	Address
	0	45098	1003	90001	Akrit	ti Singh	F	2002- 05-09	8908908398	Indian	Longterm	23- asp road kolkata 64
	1	45099	1003	90003	Pritan	n Bannerjee	М	2001- 02-02	9087988092	Indian	Short term	12- sp Mukherjee road kolkata 87
	2	45100	1003	90005	Salin	n Khan	М	2001- 06-09	9900887766	Indian	Long term	34- roy road howrah 34
	3	45101	1003	90002	Yasoo	f Akhtar	М	2000- 08-07	9988455767	Indian	Long term	34 chellam road Chennai 01
	4	45102	1003	90003	Wahida	a Begum	F	2003- 07-08	9090123467	Indian	Short term	46 bb roy road karnataka 04
[13]: [14]:	cursor=connection.cursor()											
					1000000							
[18]:	df.		op('First_n rop('Last_n)									
	df.	2=df.dro 2=df2.dr 2.head()	rop('Last_n	ame',axis	=1)	DOB F	Phone N	łationality	/ Insurance_ty	pe Addr	ess Country	Name
	df.	2=df.dro 2=df2.dr 2.head()	rop('Last_n	ame',axis	Gender	DOB F 2002- 05-09 890890		lationality Indian	5 5	23-	asp oad India\r	Name Akriti Singh
	df. df.	2=df.drd 2=df2.dr 2.head() Patient_ID	rop('Last_n) Insurance_ID	ame',axis=	Gender	2002- ₈₉₀₈₉	18398	8 08	Longter	23- rm r kolkata 12	asp oad India\r a 64 - sp rjee oad India\r	Akriti
[18]; t[18];	df. df.	2=df.drc 2=df2.dr 2.head() Patient_ID 45098	Insurance_ID	Hospital_ID	Gender F	2002- 05-09 890890	98398 98092	Indiar	Longter	23- rm r kolkata 12 Mukhe rm kolkata	asp oad India\r 664 - sp rjee oad India\r	Akriti Singh Pritam
	df. df. df.	2=df.drc 2=df2.dr 2.head() 45098	Insurance_ID 1003	Hospital_ID 90001	Gender F M M	2002- 05-09 890890 2001- 02-02 908798	08398 08392 087766	Indian	Longter Short ter Long ter	rm 23- rm kolkata 12 rm Mukhe n kolkata 34- roy n howrah 34 chel	asp oad India\r 664 -sp rjee oad India\r 0ad India\r	Akriti Singh Pritam Bannerjee

```
[1]: import pymysql
     import pandas as pd
     import sqlalchemy
[2]: conn=pymysql.
     →connect(host='localhost',port=int(3306),user='root',passwd='password',db='Hospital_DB')
[3]: df=pd.read_sql_query("SELECT * FROM Insurer",conn)
    print(df.head())
       InsuranceID
                                      CompanyName
                                                                InsuranceName
              1001 Aegon Life Insurance Co. Ltd. Individual Health Insurance
              1002 Edelweiss Life Insurance Co.
    1
                                                      Family Health Insurance
    2
              1003 Birla Sun Life Insurance Co.
                                                   Critical illness Insurance
        StartDate
                      EndDate EffectiveDate InsuranceBaseAmount InsuranceStatus \
    0 2020-01-01 2036-01-01
                                2020-01-15
                                                         200000
                                                                         Active
    1 2020-01-03 2039-01-03
                                                         300000
                                2020-03-15
                                                                         Active
    2 2020-02-05 2041-02-05
                                2020-05-17
                                                         400000
                                                                         Active
       Term
            TopupAmount InsuranceCover
    0
         15
                  200000
    1
         18
                  300000
                                       3
         20
                  150000
                                       3
[4]: cursor=conn.cursor()
[5]: data=pd.DataFrame({
         'Insurance_amount':df["InsuranceBaseAmount"]+df["TopupAmount"]
    })
[6]: data
       Insurance_amount
[6]:
                 400000
    0
                 600000
    1
    2
                 550000
```

```
39 In Process , ACTIVE
         40
                    Invalid , INVALID
         41 In Process , ACTIVE
         42
             In Process , RECOVERED
         43 In Process , ACTIVE
         44
                Settled, RECOVERED
         45 invalid , INVALID
         46
                 Settled , RECOVERED
         47 Settled , ACTIVE
         48
                 Settled , RECOVERED
         49 In Process , ACTIVE
         50
                  In Process , ACTIVE
         51 In Process , ACTIVE
         52
                   Invalid , INVALID
         53 In Process , ACTIVE REVIEW
         54
                   Invalid , INVALID
         55 Settled , RECOVERED
                   Invalid , INVALID
In [26]: CLAIMS=pd.DataFrame({'Claim': df['Claim_ID']})
In [27]: CLAIMS
Out[27]: Claim
        1 4009
         2 4014
         4 4020
        5 4024
          6 4026
        7 4030
          8 4031
         9 4037
         10 4038
         11 4042
         12 4045
         13 4048
         15 4051
         16 4053
         17 4055
         18 4059
         19 4060
         20 4061
         21 4066
         22 4069
         23 4070
         24 4072
         25 4073
         26 4074
        27 4077
         29 4081
         30 4086
         31 4089
         32 4090
         33 4096
         34 4098
         35 4100
         36 4101
         37 4103
         38 4105
         39 4107
         40 4109
         41 4110
```

```
13 2022-05-08
               14 2022-04-23
               15 2021-12-31
               16 2021-05-22
               17 2022-02-02
               18 2021-09-16
              19 2022-01-14
               20 2022-01-21
              21 2021-12-08
               22 2022-03-03
               23 2022-03-01
               24 2021-12-11
              25 2022-01-23
               26 2022-01-15
              27 2022-01-02
               28 2022-01-01
              29 2022-11-02
               30 2022-01-22
              31 2022-01-02
               32 2022-07-03
              33 2021-03-11
              35 2022-12-05
               36 2021-09-08
              37 2022-05-04
               38 2021-03-09
              39 2021-09-29
              41 2022-04-02
               42 2022-04-23
              43 2021-12-02
               44 2022-01-05
               45 2022-06-05
               46 2021-10-10
              47 2021-11-12
               48 2021-04-12
               49 2022-04-19
               50 2022-03-10
               51 2022-05-05
               52 2022-01-01
               53 2022-03-29
               54 2022-03-19
              55 2022-05-10
               56 2022-09-05
In [30] df=pd.read_sql_query("SELECT * FROW INSURER",connection)
                print(df.head())
                               ce_ID Company_name Insurance_name
1001 Aegon Life Insurance Co. Ltd. Individual Health Insurance
1002 Edelweiss Life Insurance Co. Family Health Insurance
1003 Birla Sun Life Insurance Co. Critical illness Insurance
                   Insurance_ID
                                                                                                                Insurance_name \
               0
               | Start_date | End_date Effective_date Insurance_Status | Term | Topup_amount | \ 0 | 2020-01-01 | 2036-01-01 | 2020-01-15 | 200000 | 0 | 15 | \ 1 | 2020-01-03 | 2039-01-03 | 2020-03-15 | 300000 | 0 | 18 | \ 2 | 2020-02-05 | 2041-02-05 | 2020-05-17 | 400000 | 0 | 20
                   Insurance_Cover
                                  200000
                                  300000
              /home/ubb01/anaconda3/lib/python3.9/site-packages/pandas/io/sql.py:761: UserWarning: pandas only supp
ort SQLAlchemy connectable(engine/connection) ordatabase string URI or sqlite3 DBAPI2 connectionother
DBAPI2 objects are not tested, please consider using SQLAlchemy
warnings.warn(
In [68]: df=pd.read_sql_query("SELECT * FROM INSURER",connection) print(df.head())
                                ce_ID Company_name Insurance_name \
1001 Aegon Life Insurance Co. Ltd. Individual Health Insurance \
1002 Edelweiss Life Insurance Co. Family Health Insurance \
1003 Birla Sun Life Insurance Co. Critical illness Insurance
                   Insurance ID
               0
                               1003
```

SQOOP and HIVE and HDFS

All tables are imported to HDFS from MYSQL

```
22/06/07 15:25:55 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applica
ble
Found 8 items
                                            0 2022-06-07 15:19 Claim
drwxr-xr-x
             - ubh01 supergroup
drwxr-xr-x
              - ubh01 supergroup
                                            0 2022-06-07 14:58 Final table
                                            0 2022-06-07 15:21 Hospital
0 2022-06-07 15:23 Insurer
              - ubh01 supergroup
drwxr-xr-x
              - ubh01 supergroup
drwxr-xr-x
                                            0 2022-06-07 15:22 Patient
0 2022-06-07 15:24 groups
drwxr-xr-x
             - ubh01 supergroup
              - ubh01 supergroup
drwxr-xr-x
              - ubh01 supergroup
drwxr-xr-x
                                             0 2022-06-07 15:25 subgroup
              - ubh01 supergroup
                                            0 2022-06-07 11:11 ubh01
ubh01@ubh01:~$ hdfs dfs -ls Final_table
22/06/07 15:26:32 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applica
ble
Found 2 items
-rw-r--r-- 1 ubh01 supergroup
-rw-r--r-- 1 ubh01 supergroup
                                            0 2022-06-07 14:58 Final_table/_SUCCESS
                                        10229 2022-06-07 14:58 Final_table/part-m-00000
ubh01@ubh01:~$
```

We have used Sgoop for the import from HDFS to HIVE

```
ubh01@ubh01:~$ sqoop import --connect jdbc:mysql://ubh01/project -table Final_table --fields-terminated-by '|' --username root --password passwo
 rd -m 1
Warning: /home/ubh01/sqoop-1.4.7.bin__hadoop-2.6.0/../hcatalog does not exist! HCatalog jobs will fail.
Please set $HCAT_HOME to the root of your HCatalog installation.
Warning: /home/ubh01/sqoop-1.4.7.bin__hadoop-2.6.0/../accumulo does not exist! Accumulo imports will fail.
Please set $ACCUMULO_HOME to the root of your Accumulo installation.
Warning: /home/ubh01/sqoop-1.4.7.bin__hadoop-2.6.0/../zookeeper does not exist! Accumulo imports will fail.
Please set $ZOOKEEPER_HOME to the root of your Zookeeper installation.
22.706/07 14:57:36 IMFO sqoop Sqoop: Purping Sqoop yersion: 1.4.7
22/06/07 14:57:36 INFO sqoop.Sqoop: Running Sqoop version: 1.4.7
22/06/07 14:57:36 WARN tool.BaseSqoopTool: Setting your password on the command-line is insecure. Consider using -P instead.
22/06/07 14:57:36 INFO manager.MySQLManager: Preparing to use a MySQL streaming resultset.
22/06/07 14:57:36 INFO tool.CodeGenTool: Beginning code generation
Loading class `com.mysql.jdbc.Driver'. This is deprecated. The new driver class is `com.mysql.cj.jdbc.Driver'. The driver is automatically regis tered via the SPI and manual loading of the driver class is generally unnecessary.
22/06/07 14:57:37 INFO manager.SqlManager: Executing SQL statement: SELECT t.* FROM `Final_table` AS t LIMIT 1
22/06/07 14:57:37 INFO manager.SqlManager: Executing SQL statement: SELECT t.* FROM `Final_table` AS t LIMIT 1
22/06/07 14:57:37 INFO orm.CompilationManager: HADOOP_MAPRED_HOME is /home/ubh01/hadoop-2.7.1
Note: /tmp/sqoop-ubh01/compile/bbfb98d630adc271fc0b148d42042c24/Final_table.java uses or overrides a deprecated API.
Note: Recompile with -Xlint:deprecation for details.
22/06/07 14:57:40 INFO orm.CompilationManager: Writing jar file: /tmp/sqoop-ubh01/compile/bbfb98d630adc271fc0b148d42042c24/Final_table.jar
22/06/07 14:57:40 WARN manager.MySQLManager: It looks like you are importing from mysql.
22/06/07 14:57:40 WARN manager.MySQLManager: This transfer can be faster! Use the --direct
22/06/07 14:57:40 WARN manager.MySQLManager: option to exercise a MySQL-specific fast path.
22/06/07 14:57:40 INFO manager.MySQLManager: Setting zero DATETIME behavior to convertToNull (mysql)
22/06/07 14:57:40 INFO mapreduce.ImportJobBase: Beginning import of Final_table
SLF4J: Class path contains multiple SLF4J bindings.
SLF4J: Found binding in [jar:file:/home/ubh01/hadoop-2.7.1/share/hadoop/common/lib/slf4j-log4j12-1.7.10.jar!/org/slf4j/impl/StaticLoggerBinder.c
lass]
SLF4]: Found binding in [jar:file:/home/ubh01/hbase-1.1.2/lib/slf4j-log4j12-1.7.5.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: See http://www.slf4j.org/codes.html#multiple_bindings for an explanation.
SLF4J: Actual binding is of type [org.slf4j.impl.Log4jLoggerFactory]
22/06/07 14:57:40 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applica
ble
22/06/07 14:57:40 INFO Configuration.deprecation: mapred.jar is deprecated. Instead, use mapreduce.job.jar
22/06/07 14:57:41 INFO Configuration.deprecation: mapred.map.tasks is deprecated. Instead, use mapreduce.job.maps
22/06/07 14:57:41 INFO client.RMProxy: Connecting to ResourceManager at /127.0.0.1:8032
```

Like this, we have created the schema for every table in HIVE

```
hive> use project;
Time taken: 0.038 seconds
hive> create table final_table (
> Patient_name string,Patient_address string,Claim_ID int,Insurance_details string,
       Claim_status_details string,Claim_issue_date Date,Insurer_term int,Patient_age int,Group_details string,Insurance_amount int )
       row format delimited
        fields terminated by
       location '/user/ubh01/Final_table';
Time taken: 0.413 seconds
hive> select * from final table;
Liza Kusari
                    34- ks sahani road. Mumbai 05 India
                                                                       4007
                                                                                 Critical illness Insurance, Birla Sun Life Insurance Co. In Process, ACTIVE
                                                                                                                                                                                    202
                    32 Genetic Illness,Down syndrome
34- ks sahani road. Mumbai 05 India 400
32 Accident,Road-traffic accident
                20
                                                                             550000
1-12-12
Liza Kusari
                                                                       4060
                                                                                 Critical illness Insurance, Birla Sun Life Insurance Co. In Process, ACTIVE
                                                                                                                                                                                    202
2-01-14
                                                                             550000
                                                                     ident
                    34-yk road punjab 99 India 4014 Crit
7 Critical Illness,Kidney failure 550000
Sanvi Shetty
                                                                       Critical illness Insurance, Birla Sun Life Insurance Co. Settled, RECOVERED
       20
                 27
                   34-yk road punjab 99 India 4669
7 Industrial Disease,Dermatitis
9- ss roy road Delhi 09 India 4020
                                                                       Critical illness Insurance, Birla Sun Life Insurance Co. Settled, RECOVERED
Sanvi Shetty
                                                                                                                                                                          2022-03-03
       20
                 27
                                                                     550000
Khushi Khan
                                                                       Critical illness Insurance, Birla Sun Life Insurance Co. Settled, RECOVERED
                            Accident,Burn Cases
                                                          550000
                   9- ss roy road Delhi 09 India 4073 C
Critical Illness,First heart attack
Khushi Khan
                                                                       Critical illness Insurance,Birla Sun Life Insurance Co. Settled,RECOVERED
                                                                                                                                                                          2022-01-23
                 32
       20
                                                                               550000
                              7- kk road Chennai 09 India
Accident,Current Shock
Mollika Singhania
                                                                       4026
                                                                                 Critical illness Insurance, Birla Sun Life Insurance Co. In Process, ACTIVE
                                                                                                                                                                                    202
2-05-18
                                                                   550000
Mollika Singhania
                              7- kk road Chennai 09 India
                                                                       4074
                                                                                 Critical illness Insurance, Birla Sun Life Insurance Co. In Process, ACTIVE
                                                                                                                                                                                    202
                          20
                                                                   550000
2-01-15
                20
                                    Surgeries, Nasal Concha
                   9- linkon street kolkata 9 India 4031 Critical Illness Insurance,
29 Industrial Disease,Musculoskeletal disorder 550000
29 Industrial Disease,Musculoskeletal disorder 550000
20 Industrial Disease,Musculoskeletal disorder 550000
21 Industrial Disease,Musculoskeletal disorder 550000
22 Industrial Disease,Musculoskeletal disorder 550000
                                                                                 Critical illness Insurance, Birla Sun Life Insurance Co. In Process, ACTIVE
Rohan Ghose
                                                                                                                                                                                    202
                20
2-05-14
Rohan Ghose
                                                                                                                                                                                    202
2-07-03
                20
      h Ghose 2- sr road Pune 4 India 4038 Criti
Industrial Disease,Respiratory Problems 550000
Lokesh Ghose
                                                            Critical illness Insurance, Birla Sun Life Insurance Co. Invalid, INVALID 2022-04-01 20
                                                                                                                                                                                    26
okesh Ghose
                    2- sr road Pune 4 India 4081
                                                             Critical illness Insurance, Birla Sun Life Insurance Co. In Process, ACTIVE
                                                                                                                                                                                    20
       26
                 Surgeries,Oncology
                                                550000
                              6 - Fort William street kolkata 9 India 4042
30 Accident,Burn Cases 550000
Rani Chakraborty
ACTIVE NULL
                                                                                                                                                                          In Process
                                                                                            Critical illness Insurance, Birla Sun Life Insurance Co.
                        20 30 Accident,Burn Cases 550000 ju colony kolkata 7 India 4045 Critical illness Insurance,Birla Sun Life Insurance Co. Invalid,INVALID
Afzal Alam
                                                                                                                                                                          2022-03-31
                           Genetic Illness,Trisomy 13
```

We have imported the tables: claim, final_table, groups, hospital, insurer, patient and subgroup to HIVE.

```
hive> show tables;
OK
claim
dept
dummy
dummy2
emp
final table
groups
hospital
insurer
mydevices
patient
salgrade
subgroup
webpage
webpage titanic
Time taken: 0.064 seconds, Fetched: 15 row(s)
```

Final processing from data

Analyzing data from Final table to plot the final Graphs.

Claims based on last 6 months.

```
[1]: import pymysql
     import pandas as pd
     import sqlalchemy
[2]: conn=pymysql.
     -connect(host='localhost',port=int(3306),user='root',passwd='password',db='Hospital_DB')
[3]: import matplotlib.pyplot as pl
[4]: mycursor = conn.cursor()
[5]: mycursor.execute("select Claim_issue_date,Patient_age from FinalTable where_
      →Claim_issue_date >=now()-interval 6 month;")
     result = mycursor.fetchall
[6]: IssueDate = []
     Age = []
     for i in mycursor:
        IssueDate.append(i[0])
         Age.append(i[1])
     print("Claim issue date = ", IssueDate)
     print("Patient age = ", Age)
    Claim issue date = [datetime.date(2021, 12, 12), datetime.date(2022, 5, 9),
    datetime.date(2022, 1, 3), datetime.date(2022, 4, 30), datetime.date(2022, 5,
    18), datetime.date(2022, 4, 21), datetime.date(2022, 5, 14), datetime.date(2022,
    4, 7), datetime.date(2022, 4, 1), datetime.date(2022, 3, 31),
    datetime.date(2022, 5, 8), datetime.date(2022, 4, 23), datetime.date(2021, 12,
    31), datetime.date(2022, 2, 2), datetime.date(2022, 1, 14), datetime.date(2022,
    1, 21), datetime.date(2021, 12, 8), datetime.date(2022, 3, 3),
    datetime.date(2022, 3, 1), datetime.date(2021, 12, 11), datetime.date(2022, 1,
    23), datetime.date(2022, 1, 15), datetime.date(2022, 1, 2), datetime.date(2022,
    1, 1), datetime.date(2022, 1, 22), datetime.date(2022, 1, 2),
    datetime.date(2022, 5, 4), datetime.date(2022, 2, 2), datetime.date(2022, 4, 2),
    datetime.date(2022, 4, 23), datetime.date(2022, 1, 5), datetime.date(2022, 4,
    19), datetime.date(2022, 3, 10), datetime.date(2022, 5, 5), datetime.date(2022,
    1, 1), datetime.date(2022, 3, 29), datetime.date(2022, 3, 19),
    datetime.date(2022, 5, 10)]
    Patient age = [32, 21, 27, 18, 20, 26, 29, 28, 26, 23, 27, 23, 21, 10, 32, 26,
    28, 27, 29, 27, 32, 20, 38, 39, 39, 35, 10, 28, 21, 27, 33, 36, 21, 24, 22, 23,
    35, 30]
  1: pl.rcParams['figure.figsize']=[20,8]
     pl.bar(IssueDate, Age,width=3)
     pl.xlabel("Claims based on last 6 months")
     pl.ylabel("Patient Age")
     pl.title("Claim Details")
     pl.show()
```

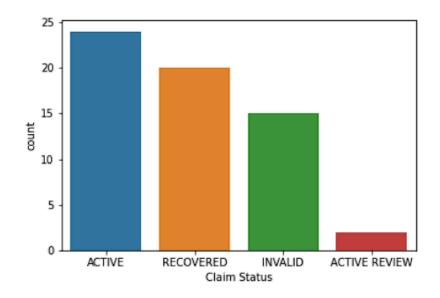
Patient claims based on disease.

```
gender_dict=_dict(zip(Patient_df.Patient_ID, Patient_df.Gender))
Claim_df["Gender"]=Claim_df.apply(lambda x:gender_dict[x[0]], axis=1)
print(Claim_df)
gendf=pd.DataFrame()
gendf=pd.pivot_table(Claim_df, index=['Group'], columns=['Gender'], values=['Patient_ID'], aggfunc='count')
print(gendf)
plt.figure()
gendf.plot.bar(title='STACKED BAR CHART', stacked=True)
plt.xlabel('Types of Group')
plt.ylabel('No.of Customers')
plt.savefig("groupgenderanalysis.png",bbox_inches='tight')
```

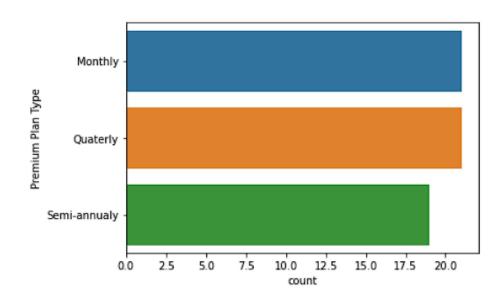
```
gdf= pd.DataFrame()
gdf=Claim_df["Group"].value_counts()
print(gdf)
plt.figure()
gdf.plot.bar(title='Patient and Group Analysis',color='red')
plt.xlabel('Types of Group')
plt.ylabel('No of patients')
plt.savefig("groupanalysis.png",bbox_inches='tight')
```

❖ OUTPUT SCREEN

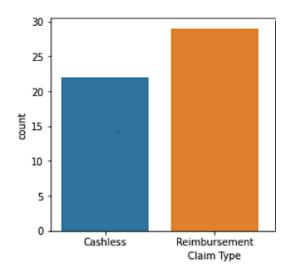
Different claims made and the claim status



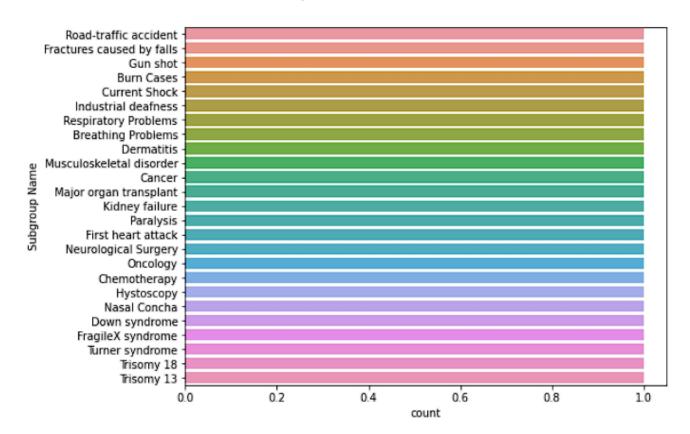
Insurance Premium Plan Type



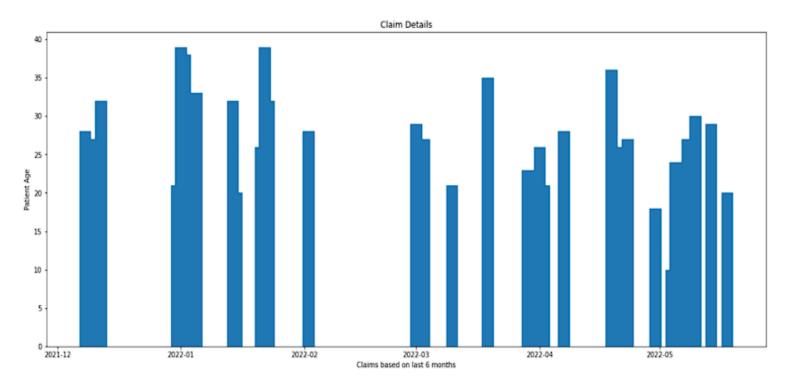
Type of claim request Cashless or Reimbursement



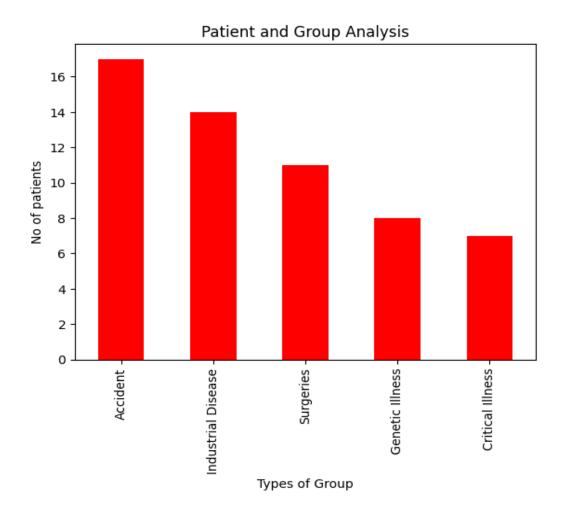
Subgroups of Disease



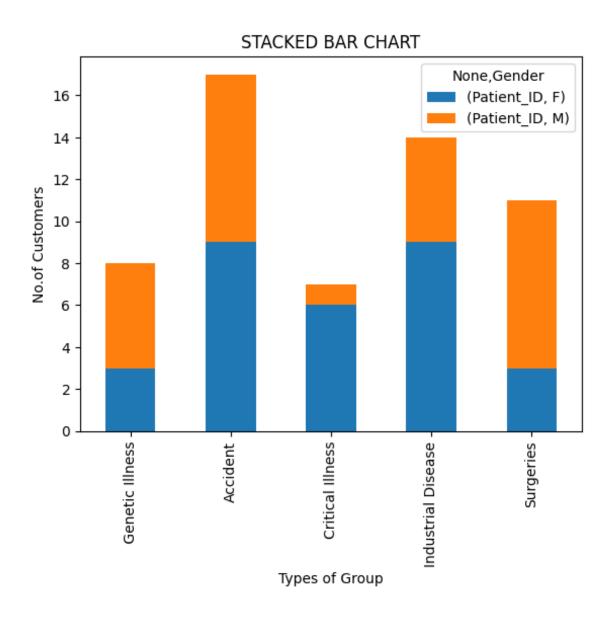
Patient claims based on past 6 months



Patient claims based on disease



Patient claims based on disease, classified by Gender



Business Benefits/Conclusion

The company is facing challenges to get the insights they need to improve their services as the dataset is on a much bigger scale and may contain missing records. So the company wants to take the help of Data Analytics using Big Data Ecosystem in order to evaluate the quality of care provided by health care providers. To do this, we have been assigned to identify patterns of past claims made for insurance in order to understand the customers better, and provide targeted experience and improve customer retention.

Further Enhancements/Recommendations

This project has a vast scope in future for all the Insurance Companies. Companies can get detailed analysis of data from this project and can serve the customers according to their special needs and benefits. They can also change their insurance plans and business needs for maximum profits and also provide with the optimal service to patients.

❖ References/Bibliography

- i. Python Data analytics
- ii. DrawlO
- iii. GeeksforGeeks
- iv. Hive
- v. Jupyter
- vi. Hadoop

❖ GitHub Link

BigData-Project-Hospital-Claim-Management