Pandemic Information System Model

SYSTEMS AND METHODS FOR BIG AND UNSTRUCTURED DATA

PROF. MARCO BRAMBILLA

THIRD DELIVERY
ELASTICSEARCH PROJECT

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Avci Oguzhan - 10557284 Gentile Nicole - 10594355 Rigamonti Davide - 10629791 Singh Raul - 10623232 Tagliaferri Mattia - 10572418



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1 Introduction

1.1 Problem Specification

The idea of the project is to use Elasticsearch to store and analyze data about COVID-19 vaccinations in Italy. The data we used can be found in the following GitHub repository: https://raw.githubusercontent.com/italia/covid19-opendata-vaccini/master/dati/somministrazioni-vaccini-latest.csv and contains data about daily vaccinations divided by region, age range and vaccine supplier.

Moreover, we used three additional datasets to integrate other information about the population of each region (*platea.csv*), the number of healed people in each region (*soggetti-guariti.csv*) and data about the number of daily covid cases (*dpc-covid19-ita-regioni.csv*).

The main goal is to analyze the data in order to build meaningful statistics about the vaccinations in Italy; examples can be the total number of doses for each vaccine brand, the top regions for vaccine administration and so on.

1.2 Hypoteses

We assumed that the general user doesn't know the italian language, therefore all the fields' names contained in the dataset were renamed to an appropriate english translation.

For the file dpc-covid19-ita-regioni.csv we decided to not include some fields mainly because they are no longer updated or they are not entirely relevant for the scope of this project.

2 Data

2.1 Schema description

- Schema for vaccines administrations (somministrazioni-vaccini-latest.csv):
 - @timestamp: date
 Timestamp assigned by elasticsearch.
 - administration_date (data_somministrazione): date [iso8601]
 Date of all the doses' administrations using the international standard format.
 - supplier (fornitore): keyword
 Brand of the administrated vaccine, mapped as keyword since entries are enumerated and aggregation is possible.
 - area (area): keyword
 Abbreviation for a region's name, mapped as keyword since entries are enumerated and aggregation is possible.
 - age_range (fascia_anagrafica): keyword
 Age range of the people administered with the vaccines, mapped as keyword since entries are enumerated and aggregation is possible.
 - male_gender (sesso_maschile): long
 Total number of vaccinations administered to males by day and region, mapped as a long since it represents a possibly large number.
 - female_gender (sesso_femminile): long
 Total number of vaccinations administered to females by day and region, mapped as a long since it represents a possibly large number.
 - first_dose (prima_dose): long
 Total number of administered first doses, mapped as a long since it represents a possibly large number.
 - second_dose (seconda_dose): long
 Total number of administered second doses, mapped as a long since it represents a possibly large number.
 - previous_infection (pregressa_infezione): long
 Total number of administered doses to subjects with a previous Covid-19 infection within 3 to 6 months before the administration date, therefore needing only one dose, mapped as a long since it represents a possibly large number.
 - booster_dose (dose_addizionale_booster): long
 Total number of administered booster doses, mapped as a long since it represents a possibly large number.
 - total_doses: long
 Total number of administered doses considering first, second, booster and previous infection doses, mapped as a long since it represents a possibly large number.

NUTS1_code (codice_NUTS1): keyword

European classifications for territorial units (Level: NUTS 1), mapped as keyword since entries are enumerated and aggregation is possible.

- NUTS2_code (codice_NUTS2): keyword

European classifications for territorial units (Level: NUTS 2), mapped as keyword since entries are enumerated and aggregation is possible.

ISTAT_region_code (codice_regione_ISTAT): keyword

ISTAT code used for identifying regions, mapped as keyword since entries are enumerated and aggregation is possible. Additionally, in order to fully utilize the map representation capabilities of Kibana this value needs to be joined with Kibana's own ISTAT regional codes (stored as keywords), besides, it would be useless to execute arithmetical operations on this value due to its nature.

- area_name (nome_area): text

Full name of the region mapped as text in order to be better suited for queries that utilize an analyzer. Even though this field technically represents an enumeration, we found particulary useful the ability to search for a region's name utilizing a text analyzer since there are plenty other fields that can be used to aggregate regions.

• Schema for **population** (platea.csv):

- area (area): keyword

Abbreviation for a region's name, mapped as keyword since entries are enumerated and aggregation is possible.

- area_name (nome_area): text

Full name of the region mapped as text in order to be better suited for queries that utilize an analyzer.

- age_range (fascia_anagrafica): keyword

Age range of the people administered with the vaccines, mapped as keyword since entries are enumerated and aggregation is possible.

- total_population (totale_populatione): long

Total population for an associated area and age range, mapped as a long since it represents a large number.

• Schema for **healed** (soggetti-guariti.csv):

- area (area): keyword

Abbreviation for a region's name, mapped as keyword since entries are enumerated and aggregation is possible.

- area_name (nome_area): text

Full name of the region mapped as text in order to be better suited for queries that utilize an analyzer.

age_range (fascia_anagrafica): keyword

Age range of the people administered with the vaccines, mapped as keyword since entries are enumerated and aggregation is possible.

total_healed (totale_healed): long

Total number of people that heald from an infection in a 6 months timespan for an associated area and age range, mapped as a long since it represents a large number.

• Schema for **covid-information** (dpc-covid19-ita-regioni.csv):

- @timestamp: date

Timestamp assigned by elasticsearch.

- date: date [iso8601]

Date of the government information.

area_name (denominazione_regione): text

Full name of the region mapped as text in order to be better suited for queries that utilize an analyzer.

ISTAT_region_code: keyword

ISTAT code used for identifying regions, mapped as keyword since entries are enumerated and aggregation is possible.

- deceased: long

Total number of people that are deceased due to Covid-19, mapped as a long since it represents a large number.

- recovered: long

Total number of people that have been dismissed from the hospital because they have recovered from Covid-19, mapped as a long since it represents a large number.

- home_confinement: long

Total number of people that are home confined due to Covid-19, mapped as a long since it represents a large number.

- lat: double

Latitude of the region, mapped as a double since it's a float number.

- long: double

Longitude of the region, mapped as a double since it's a float number.

- notes: text

General notes from the government, mapped as text in order to be better suited for queries that utilize an analyzer.

- cases_notes: text

Notes related to the tested cases from the government, mapped as text in order to be better suited for queries that utilize an analyzer.

- new_positives: long

Daily new number of people that have been tested positive due to Covid-19, mapped as a long since it represents a large number.

hospitalized_with_symptoms: long

Total number of people that have been hospitalized with symptoms of Covid-19, mapped as a long since it represents a large number.

- country: keyword

Code indicating the country, mapped as keyword since entries are enumerated and aggregation is possible.

- tests: *long*

Total number of people that have been tested for Covid-19 (using molecular tests), mapped as a long since it represents a large number.

- intensive_care: long

Total number of people that are in intensive care due to Covid-19, mapped as a long since it represents a large number.

- total_cases: long

Total amount of positive cases, mapped as a long since it represents a large number.

- total_hospitalized: long

Total number of people that have been hospitalized due to Covid-19, mapped as a long since it represents a large number.

total_positives: long

Total number of people that have been found positive to Covid-19 (either via test or home confinement), mapped as a long since it represents a large number.

total_positives_variation: long

Daily variation of the total number of people that have been tested positive to Covid-19, mapped as a long since it represents a large number.

- location: geo_point

The combination of the latitude and longitude fields, mapped as geo_point since they represent a set of coordinates.

2.2 Dataset description

The dataset contains data about COVID-19 vaccinations in Italy.

For each day, region, age range and vaccine supplier we have information about: the number of vaccinated people (male and female), the number of 1st, 2nd and booster doses and information about the number of people who was previously infected.

We decided to integrate this data using the 'platea.csv' dataset that can be found in the specified repository; it contains data about the population for each region and each age range.

We also used information about the number of healed people in each region that can be found in 'soggetti-guariti.csv' present in the same repository.

In addition, we used the 'dpc-covid19-ita-regioni.csv' dataset to have more information about the pandemic, from the total number of positives to the number of hospitalized and deceased people.

2.3 Creating indexes

The four indexes utilized throughout the project were created following a specific naming convention.

We created one index for each CSV file (*somministrazioni-vaccini-latest.csv*, *platea.csv*, *soggetti-guariti.csv*, *dpc-covid19-ita-regioni.csv*), and named them respectively: "index-vaccines_administrations", "index-population", "index-healed" and "index-covid-information".

Then, we added in the "index-patterns" section an **aggregate index** named "index" as to create an overall "index-*" that combines all the indexes allowing for complex operations on multiple indexes.

2.4 Queries

2.4.1 Number of daily vaccines for each dose

Total number of first, second and booster doses administred in each date.

```
GET /index-vaccines_administrations/_search
"size": 0,
  "aggs": {
    "Administration Date": {
      "terms": {
        "field": "administration_date",
        "size": 370
      "aggs": {
         'totalFirstDoses": {
          "sum": {
            "field": "first_dose"
        "totalSecondDoses": {
          "sum": {
            "field": "second_dose"
        "totalBoosters": {
          "sum": {
            "field": "booster_dose"
```

2.4.2 Number of doses per gender in each region

Total doses administred to men and to women in each region.

2.4.3 Number of vaccines per dose number in each region

Total number of first, second, booster and total doses administred in each region.

```
GET /index-vaccines_administrations/_search
"size": 0,
  "aggs": {
    "Region": {
      "terms": {
       "field": "area",
        "size": 21
      "aggs": {
        "totalFirstDoses": {
          "sum": {
            "field": "first_dose"
        "totalSecondDoses": {
          "sum": {
           "field": "second_dose"
        "totalBoosters": {
          "sum": {
           "field": "booster_dose"
        "totalDoses": {
          "sum": {
           "field": "total_doses"
    }
 }
```

2.4.4 Number of doses for each vaccine supplier

Total doses for each vaccine supplier (Pfizer/BioNTech, Moderna, Vaxzevria (AstraZeneca), Janssen, Pfizer Pediatrico).

2.4.5 Number of doses for each NUTS1 zone

Total number of doses administred in each NUTS1 zone (ITC= NorthWest, ITH = NorthEast, ITI = Center, ITF = South, ITG = Isles)

2.4.6 Number of people with previous infection by region

Total number of people who were infected by Covid-19 within 3 to 6 months before the date of the vaccination for each region.

2.4.7 Number of doses for each age group

Total number of doses administred to each age range ('05-11', '12-19', '20-29', '30-39', '40-49', '50-59', '60-69', '70-79', '80-89', '90+').

2.4.7.1 Grouping by age range and region

Total number of doses administred to each age range in each region.

2.4.8 Number of average doses per region

Average number of administred doses in each region.

```
GET /index-vaccines_administrations/_search
{"size":0,
    "aggs": {
        "terms": {
            "field": "area"
        },
        "aggs": {
            "avg_doses": { "avg": {
                "field": "total_doses"
            }        }
        }
     }
}
```

2.4.9 Number of doses for each age group and vaccine supplier

Total number of doses by age group and vaccine supplier.

```
GET /index-vaccines_administrations/_search
  "size": 0,
  "track_total_hits": true,
  "aggs": {
    "age_groups_types": {
      "composite": {
        "sources": [{
          "fascia_anagrafica": {
            "terms": {
              "field": "age_range"
          }
        }, {
          "supplier": {
            "terms": {
              "field": "supplier"
        }],
        "size": 10000
      "aggs": {
        "total_vaccines": {
    "sum": {
            "field": "total_doses"
         }
   }
```

2.4.10 Number of booster doses for age range

Total number of booster doses for each age range.

2.4.11 Top 10 dates with most vaccinations

Top 10 dates having the highest number of administred doses.

2.4.12 Search by region name

Returns information about vaccines in the specified region.

2.4.13 Percentage of fully vaccinated people per region

Returns the percentage of fully vaccinated people (i.e. having 1st and 2nd dose) for each region.

In order to obtain this resutl, we used additional data about the total population of each region contained into *platea.csv*.

```
GET /index-vaccines_administrations,index-population/_search
  "size": 0,
  "fields": [
      "field": "@timestamp",
"format": "date_time"
      "field": "administration_date",
      "format": "date_time"
  "aggs": {
    "area_percentage": {
      "terms": {
        "field": "area",
        "size": 26
      "aggs": {
         "hits": {
           "top_hits": {
             "size": 1,
             "_source": {
               "includes": "area_name"
         "total_population": {
           "sum": {
             "field": "total_population"
         "second_dose": {
           "sum": {
             "field": "second_dose"
         "percentage_second_doses": {
           "bucket_script": {
             "buckets_path": {
               "total_pop": "total_population",
    "second_dose": "second_dose"
             "script": "100 * params.second_dose / params.total_pop"
```

2.4.14 Percentage of healed people per region

Returns the number of healed people over the total population for each region. We used:

- Data about the total population in each region from *platea.csv*;
- The number of healed people for each region that can be found in *soggetti-guariti.csv*.

```
GET /index-healed,index-population/_search
  "size": 0,
  "fields": [
      "field": "@timestamp",
      "format": "date_time"
      "field": "administration_date",
      "format": "date_time"
 ],
"aggs": {
    "area_percentage": {
      "terms": {
    "field": "area",
        "size": 26
      "aggs": {
        "hits": {
          "top_hits": {
            "size": 1,
             "_source": {
               "includes": "area_name"
        "total_population": {
          "sum": {
            "field": "total_population"
        "healed": {
          "sum": {
            "field": "total_healed"
        },
        "percentage_healed": {
           "bucket_script": {
             "buckets_path": {
              "total_pop": "total_population",
"healed": "healed"
             "script": "100 * params.healed / params.total_pop"
        }
```

2.4.15 Daily number of total cases, hospitalizations and positives

Returns, the total number of cases, hopitalizations and positives for each day. Given that this is not the total for a single day, but the overall total that is increased day by day.

We used additional data from dpc-covid-19-ita-regioni.csv.

```
GET /index-covid-information/_search
"size": 0,
 "aggs": {
    "date": {
      "terms": {
        "field": "data",
        "size": 365,
        "order": {
          "_key": "asc"
      "aggs": {
        "total_cases": {
          "sum": {
            "field": "total_cases"
        },
"total_positives": {
          "sum": {
            "field": "total_positives"
        "total_hospitalized": {
          "sum": {
            "field": "total_hospitalized"
   }
```

2.4.16 Ratio between positive and hospitalized people

Returns the ratio between positive and hospitalized people for each region. It is calculated using the data of *dpc-covid19-ita-regioni.csv*.

```
GET /index-vaccines_administrations/_search
  "size": 0,
  "fields": [
      "field": "data",
      "format": "date"
   }
  "sort": [
   {
     "data": {
        "order": "desc"
  ],
  "aggs": {
    "ratio_positive_hospitalized": {
      "terms": {
       "field": "data",
"size": 365
      "aggs": {
        "region": {
          "terms": {
            "field": "ISTAT_region_code"
          },
"aggs": {
            "total_positives": {
              "sum": {
                "field": "total_positives"
              }
            "total_hospitalized": {
              "sum": {
                "field": "total_hospitalized"
            "percentage_positive_hospitalized": {
              "bucket_script": {
                "buckets_path": {
                  "total_hospitalized": "total_hospitalized",
                  "total_positives": "total_positives"
                "script": "100*params.total_hospitalized/params.total_positives"
         }
       }
   }
 }
```

2.5 Commands

2.5.1 Add entries for a given day

Adds a new document in the dataset.

```
POST /index-vaccines_administrations/_doc/

{
    "area" : "FVG",
    "area_name" : "Friuli-Venezia Giulia",
    "second_dose" : 0,
    "previous_infection" : 1,
    "administration_date" : "2022-01-03",
    "NUTS1_code" : "ITH",
    "ISTAT_region_code" : 6,
    "male_gender" : 3,
    "first_dose" : 2,
    "age_range" : "30-39",
    "supplier" : "Moderna",
    "total_doses" : 3,
    "booster_dose" : 0,
    "NUTS2_code" : "ITH4",
    "female_gender" : 0,
    "total_doses" : 3
}
```

2.5.2 Remove entries from a given day

Removes all documents from the index for the specified date.

```
POST /index-vaccines_administrations/_delete_by_query

{
    "query": {
        "match": {
            "administration_date": "2022-01-03"
        }
    }
}
```

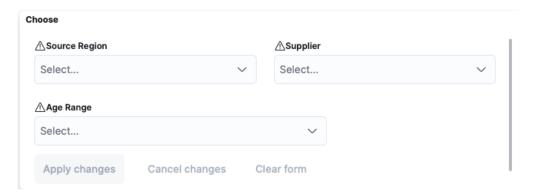
3 Data Visualization with Kibana

We used the dashboard to give a visual representation of the most meaningful queries. In order to use properly the dashboard, all imported indexes must follow the naming convention specified in the Creating indexes section.

3.1 Dashboard visualizations

3.1.1 Control panel

It allows to filter the results of the other elements of the dashboard based on region, vaccine supplier and age range.



3.1.2 Vaccinations for each region

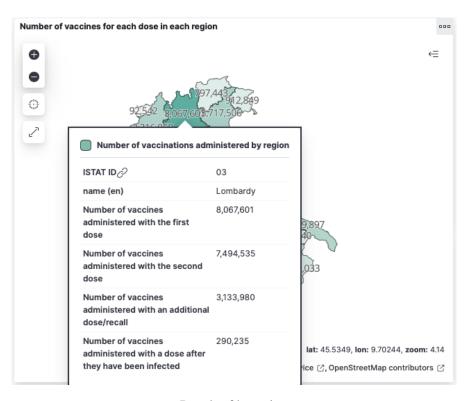
It allows to represent a dynamic map of Italy divided in regions. By default it shows the number of administred first doses in each region.

By hovering over a region, it shows a description of the number of vaccinations for each type (first, second, booster doses and so on) administered by that region.

Number of vaccines for each dose in each region



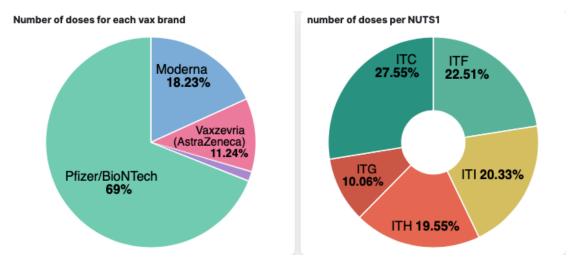
Map of Italy



Result of hovering

3.1.3 Pie charts

The two pie charts represent the percentage of doses for each vaccine supplier and for each NUTS1 over the total administred vaccines.



Percentage of doses for each vaccine brand (left) and for each NUTS1 (right)

3.1.4 Metric Visualization

It represents the percentage of people with at least one, two or three doses together with the number of people who healed from Covid-19 in the selected region (or the whole Italy if none is selected).

These numbers are obtained by integrating two different datasets:

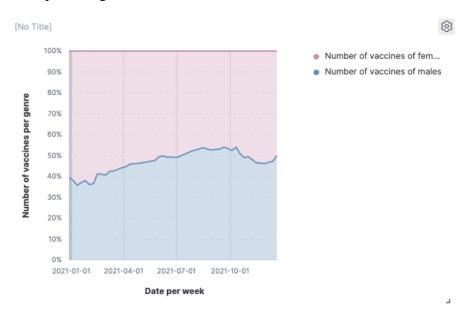
- The percentages of vaccinated people are calculated using data about the total population in each region from *platea.csv*;
- The number of healed people for each region can be found in *soggetti-guariti.csv*.



Metric visualization

3.1.5 Number of vaccines by genre

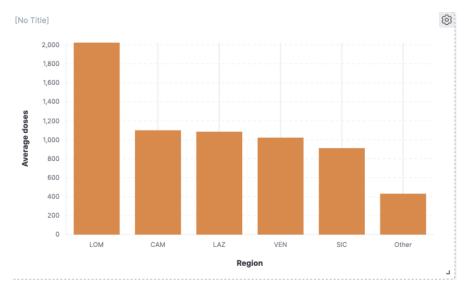
Represents a comparison between the total number of vaccines done by males and females, in each week from the starting date selected. By hovering with the mouse, a more detailed percentage is shown.



Number of vaccines by genre

3.1.6 Average number of vaccines by region

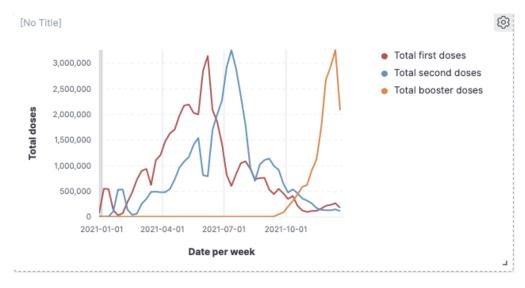
Represents the average number of vaccines administred in each region.



Average number of vaccines by region

3.1.7 Number of 1st, 2nd and booster doses per week

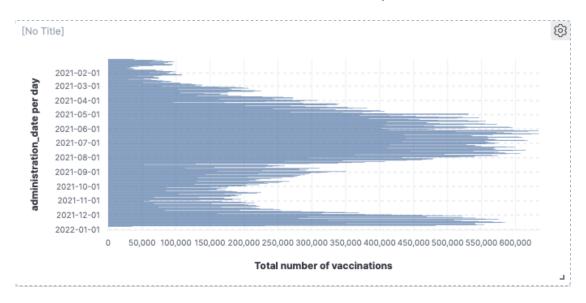
It shows a graph that, for each week from the starting date selected, represents the trend of the total doses administered by type (1st, 2nd or booster doses).



Number of 1st, 2nd and booster doses

3.1.8 Total number of vaccines by date

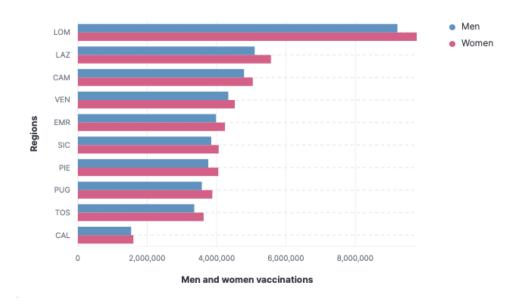
It shows the total number of administred doses in each day.



Total number of vaccines by date

3.1.9 Number of vaccines per gender in each region

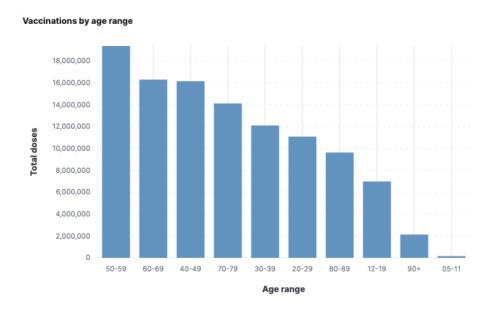
It shows the top 10 regions for number of administred doses divided by gender.



Number of vaccines per gender in each region

3.1.10 Number of vaccinations by age range

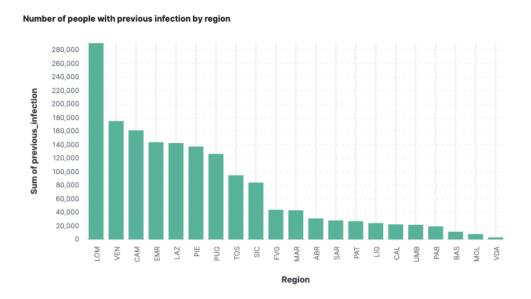
It shows the number of administred doses for each age range, sorted by number of doses.



Number of vaccinations by age range

3.1.11 Number of people with previous infection in each region

It shows, for each region, the number of people who were infected by Covid-19 before doing the vaccination.



Number of people with previous infection in each region

4 Other features in HBase

We deciced to use *HBase* as an alternative NoSQL DBMS in order to implement some additional features.

HBase is a column-oriented DBMS usually utilized in the context of the *Apache Hadoop* framework, it can be run locally using *Apache Zookeeper* or by installing *Hadoop* and running *HBase* on top of *HDFS*.

4.1 Input Data Preparation

In order to import data into *HBase*, the first step is to eliminate the labels from the csv file containing raw data.

In our case the file has been renamed as *vaccines_administrations_latest.csv* and a unique ID generated using *Python* uuid.uuid4() function (from the uuid library) has been added as the first item for each row since HBase expects a row key in that particular position.

The unique ID was generated in this way since an incremental sequential number as a row key is not recommended according to the *HBase* documentation.

Once we have the data in CSV format, we have to store it in a path from where *HBase* can access it.

If we want to load the file using Hadoop, we will have to copy the file to the *HDFS* location, this can be done using the command:

```
hadoop fs -copyFromLocal <LOCAL_PATH> <HDFS_PATH>
```

4.2 Storing data into HBase

```
create 'vaccines_administrations', {NAME => '_date'},
{NAME => '_type'}, {NAME => 'region'},
{NAME => 'stats'}, {NAME => 'doses'}, {NAME => 'nuts'}
```

This command will create an *HBase* table in order to store the data.

Notice that we are creating five column families for some of the columns of the CSV file, this is done for accessing and querying the data in a more easier and flexible way using *Apache Drill*.

```
./hbase org.apache.hadoop.hbase.mapreduce.ImportTsv
-Dimporttsv.separator=','
-Dimporttsv.columns=HBASE_ROW_KEY,_date:administration_date,
_type:supplier,region:area,stats:age_range,stats:male_gender,
stats:female_gender,doses:first_dose,doses:second_dose,
doses:previous_infection,doses:booster_dose,nuts:NUTS1_code,
nuts:NUTS2_code,region:ISTAT_region_code,region:area_name
vaccines_administrations
<PATH>
```

Once we submit this command a *MapReduce* that will import the CSV file into the HBase table that we created will start.

Each column of the CSV file is mapped into the respective column family.

<PATH> needs be changed to the HDFS path if we are using *Hadoop* or to the local path of the CSV file if we are using *Zookeeper*.

The output should resemble the following:

```
2022-01-02 11:46:09,146 INFO
2022-01-02 11:46:09,147 INFO
2022-01-02 11:46:09,157 INFO
File System Counters
                                        [main] mapreduce.Job: map 100% reduce 0%
[main] mapreduce.Job: Job job_local1490239975_0001 completed successfully
[main] mapreduce.Job: Counters: 16
                     FILE: Number of bytes read=49849936
                     FILE: Number of bytes written=31811946
                     FILE: Number of read operations=0
                     FILE: Number of large read operations=0
FILE: Number of write operations=0
          Map-Reduce Framework
                     Map input records=166273
                     Map output records=166273
Input split bytes=133
                     Spilled Records=0
                     Failed Shuffles=0
                     Merged Map outputs=0
                     GC time elapsed (ms)=80
                     Total committed heap usage (bytes)=304611328
          ImportTsv
                     Bad Lines=0
          File Input Format Counters
                     Bytes Read=18748380
          File Output Format Counters
                     Bytes Written=0
```

4.3 Querying HBase

For querying *HBase* we decided to use *Apache Drill*, this means that our queries will be written in SQL and can benefit from all the extensions that the framework provides. To be able to query an *HBase* database, after running *Apache Drill* in embedded mode, we will have to go to the storage section and enable *HBase*.

The translation of some Elasticsearch queries with sample screenshots of possible results will be provided here, a more complete selection of queries will be provided separately.

4.3.1 Number of daily vaccines for each dose

```
SELECT CONVERT_FROM(vaccines_administrations._date.administration_date, 'UTF8')

AS date_time,

SUM(CAST(vaccines_administrations.doses.first_dose AS INT))

AS number_of_daily_vaccines_first_dose,

SUM(CAST(vaccines_administrations.doses.second_dose AS INT))

AS number_of_daily_vaccines_second_dose,

SUM(CAST(vaccines_administrations.doses.booster_dose AS INT))

AS number_of_daily_vaccines_booster_dose

FROM vaccines_administrations

GROUP BY date_time
```

Possible result:

date_time \$	number_of_daily_vaccines_first_dose \$\\$\\$\$	number_of_daily_vaccines_second_dose \$\displaintarrow\$	number_of_daily_vaccines_booster_dose \$\displaystyle{\pi}\$
2021-08-12	175183	153335	0
2021-05-16	267611	137130	0
2021-10-20	35488	71043	77925
2021-01-03	35809	0	0
2021-09-21	82431	158470	8714
2021-07-08	110206	469998	0
2021-11-18	20622	28908	192862
2021-10-11	55483	82102	40970
2021-06-10	494184	128531	0
2021-04-16	275398	99764	0
Showing 1 to 10 o	of 10 entries		Previous 1 Next

4.3.2 Number of doses for each vaccine supplier

```
SELECT CONVERT_FROM(vaccines_administrations._type.supplier, 'UTF8')

AS supplier,

SUM(CAST(vaccines_administrations.doses.first_dose AS INT) +

CAST(vaccines_administrations.doses.second_dose AS INT) +

CAST(vaccines_administrations.doses.booster_dose AS INT) +

CAST(vaccines_administrations.doses.previous_infection AS INT))

AS total_doses

FROM vaccines_administrations

GROUP BY supplier
```

Possible result:

supplier	total_doses
Moderna	19718391
Vaxzevria (AstraZeneca)	12161093
Pfizer/BioNTech	74688059
Pfizer Pediatrico	158884
Janssen	1500055

4.3.3 Top 10 dates with most vaccinations

```
SELECT CONVERT_FROM(vaccines_administrations._date.administration_date, 'UTF8')

AS administration_date,

SUM(CAST(vaccines_administrations.doses.first_dose AS INT) +

CAST(vaccines_administrations.doses.second_dose AS INT) +

CAST(vaccines_administrations.doses.booster_dose AS INT) +

CAST(vaccines_administrations.doses.previous_infection AS INT))

AS total_vaccinations

FROM vaccines_administrations

GROUP BY administration_date

ORDER BY total_vaccinations ASC

LIMIT 10;
```

Possible result:

Number of vaccinations	∨ Date per day	~
44,212	2021-06-03	
38,135	2021-07-08	
37,704	2021-06-04	
37,542	2021-06-02	
33,275	2021-07-09	
33,101	2021-06-17	
32,815	2021-07-07	

5 References and sources

In order to develop this project, the following tools were used:

- Elasticsearch and Kibana in order to store, query and visualize data;
- HBase, HDFS, Hadoop, Apache Drill and Zookeeper as an alternative NoSQL framework for some features;
- Python as a mean to easily modify and prepare files for import;
- LATEX to write the report;
- Github as a versioning and collaboration mean;
- https://github.com/italia/covid19-opendata-vaccini as a source of updated and real data about vaccinations in Italy.
- https://github.com/pcm-dpc/COVID-19 as an alternative source for additional data about Covid-19 in Italy.