

# Week 6: File ingestion and schema validation

Name: Bao Khanh Nguyen

**Batch Code: LISUM04** 

**Submission Data: 10/31/2021** 

**Submitted to: Data Glacier** 

## Introduction

Validation of the schema and the ingestion of files in data pipelining is an important component of data engineering; We need to know how to read data in the most efficient way possible to maximize resource utilization. This project will be used to demonstrate different approaches of reading data and determining the best way to do so. After that, try validating input data using a YAML file and, if the file is correct, writing it to a.gz file.

### **Dataset**

The dataset that I used is "Parking\_Violations\_Issued\_-\_Fiscal\_Year\_2016"

<u>Dataset</u>. The below features are the features of this dataset. The size of the dataset is 2.15 GB.

• Summons Number: Number

• Plate ID: Plain Text

• Registration State: Plain Text

Plate Type: Plain TextIssue Date: Date & TimeViolation Code: Number

• Vehicle Body Type: Plain Text

Vehicle Make: Plain Text Issuing Agency: Plain Text

Street Code1: NumberStreet Code2: NumberStreet Code3: Number

Vehicle Expiration Date: Number
Violation Location: Plain Text
Violation Precinct: Number
Issuer Precinct: Number

• Issuer Code: Number

• Issuer Command: Plain Text

Issuer Squad: Plain Text
Violation Time: Plain Text
Time First Observed: Plain Text
Violation County: Plain Text

• Violation In Front Of Or Opposite: Plain Text

House Number: Plain Text
Street Name: Plain Text
Intersecting Street: Plain Text
Date First Observed: Number

Law Section: NumberSub Division: Plain Text

• Violation Legal Code: Plain Text

Days Parking In Effect: Plain Text
From Hours In Effect: Plain Text
To Hours In Effect: Plain Text

• Vehicle Color: Plain Text

• Unregistered Vehicle?: Plain Text

Vehicle Year: Number Meter Number: Plain Text Feet From Curb: Number

Violation Post Code: Plain Text Violation Description: Plain Text

• No Standing or Stopping Violation: Plain Text

• Hydrant Violation: Plain Text

• Double Parking Violation: Plain Text

Latitude: NumberLongitude: Number

Community Board: NumberCommunity Council: Number

• Census Tract: Number

BIN: NumberBBL: NumberNTA: Plain Text

## Solution to read this dataset

Here, I have tried to put different solutions (libraries) to the test, and the results are shown in the sections below with Wall time is the time for each method execution:

#### **Pandas:**

# Pandas option without chunk ¶

#### **Pandas with Chunks:**

# Pandas option with Chunk extension ¶

```
%%time
chunks = pd.read_csv("Parking_2016.csv", chunksize=100000)
df_pandas_chunks = pd.concat(chunks)
df_pandas_chunks.head()
```

Wall time: 1min 3s

#### Dask:

#### **Dask solution**

Wall time: 1.32 s

#### **Pyspark:**

#### Pyspark Solution

#### **CSV Disk Reader:**

#### **CSV Dict Reader Solution**

```
%%time

df_csv = csv.DictReader(open("Parking_2016.csv"))
i=0
for row in df_csv:
    print(row)
    i += 1
    if i == 5:
        break

...

ensus Tract': '', 'BIN': '', 'BBL': '', 'NTA': ''}
Wall time: 999 μs
```

#### **Datatable Fread:**

## Database, Fread Solution

```
%%time
df_dt = dt.fread("Parking_2016.csv")
df_dt.head()
Wall time: 6.95 s
```

Method	Pandas	Pandas with Chunk	Dask	Pyspark	CSV Disk Reader	Datatable Fread
Time Execution	1 min 9s	1 min 3s	1.32s	20.4s	999 x 10 <sup>-</sup> seconds o r 99 micros econds	6.95s

**In conclusion,** based on the executed times, totally "csv.DictReader" is the best one, but "dask" is also another good choice in this scenario to read data in Data frame mode. However, Dask still have some situation with dytpes, which requires we have to force dtypes for each variable, so that I will use pandas because of its stable.

# YAML File

Firstly, we have to create config file to store information about the dataset:

```
input:
format: csv
name: Parking 2016
delimiter: ","
columns:
 - Summons Number
 - Plate ID
 - Registration State
  - Plate Type
 - Issue Date
  - Violation Code
 - Vehicle Body Type
 - Vehicle Make
 - Issuing Agency
  - Street Code1
 - Street Code2
  - Street Code3
 - Vehicle Expiration Date
 - Violation Location
  - Violation Precinct
  - Issuer Precinct
  - Issuer Code
  - Issuer Command
  - Issuer Squad
  - Violation Time
 - Time First Observed
  - Violation County
 - Violation In Front Of Or Opposite
 - House Number
  - Street Name
 - Intersecting Street
  - Date First Observed
 - Law Section
  - Sub Division
 - Violation Legal Code
 - Days Parking In Effect
```

```
- From Hours In Effect
 - Vehicle Color
 - Unregistered Vehicle?
 - Vehicle Year
 - Meter Number
 - Violation Post Code
 - Violation Description
 - No Standing or Stopping Violation
 - Hydrant Violation
 - Double Parking Violation
 - Longitude
 - Community Council
 - Census Tract
 - BBL
name: trusted_Parking_2016
delimiter: "|"
path: C:\Users\nguye\Desktop\Data Science Stuff\Projects\Python\File ingestion and schema validation\
```

All of information about input and output data, and columns also be included in this file

We retrieve data and check all validations, such as column counts and names, in the following code, and if everything checks out, we may write the file in.gz format as an output:

```
: def validate_data(raw_df):
      raw_columns = list(map(lambda x: x.lower(), cfg.input.columns))
raw_columns = list(map(lambda x: x.lower(), raw_df.columns))
      trusted_columns = [x.strip(' ') for x in trusted_columns]
raw_columns = [x.strip(' ') for x in raw_columns]
           if len(raw_columns)!=len(trusted_columns):
                print(f'Count of columns are invalid! It should be {len(trusted_columns)}, but it is {len(raw_columns)}')
                return
           if raw_columns.sort()!=trusted_columns.sort():
                print('Columns are invalid!')
                return
           if raw columns.sort()!=trusted_columns.sort():
               print(f'Columns are invalid!')
                print(f'Columns in Uploaded Dataset: {list(set(raw columns).difference(trusted columns))} VS. Columns in Config File
                return
           output_file = cfg.output.name+"."+cfg.output.format
           output_path = cfg.output.path+output_file
           df.to_csv(output_path, header=None, index=None, sep=cfg.output.delimiter, compression='infer')
           print(f'File is uploaded successfully and written to: {output_file}')
           return
: validate_data(df)
```

Information about this file also be can be found:

File is uploaded successfully and written to: trusted\_Parking\_2016.gz

df = read\_data\_summary(cfg)

The size of the file is: 2151937808 Bytes It has: 51 Columns and 10626899 Rows