# ASA (Atractiveness Score Analysis)

June 30, 2022

# 1 Mobile Communication Companies in Georgia ranked by Attractiveness Score

Lets explore which mobile communication companies are operating in Georgia, how many subscribers do each of them serve, and how attractive each of them is.

Here an attractiveness of a company, or more specifically the "Attractiveness Score" which is defined later in this document, measures an ability of company to attract and retain subscribers.

Finally, best performing companies according to the "Attractiveness Score", for each year we have data for, and per category of subscribers (categories being "All subscribers" combined, "Legal Entity subscribers" and "Natural Person subscribers"), are found via this analysis and are listed towards the end of this notebook.

```
[1]: from datetime import datetime, timedelta
nb_st = datetime.utcnow()
print(f"\nNotebook START time: {nb_st} UTC\n")
```

Notebook START time: 2022-06-30 01:43:31.170139 UTC

```
[2]: %%HTML
     <style>
     @media (max-width: 540px) {
       .output .output_subarea {
         max-width: 100%;
       }
     }
     </style>
     <script>
       function code_toggle() {
         if (code shown){
           $('div.input').hide('500');
           $('#toggleButton').val(' Show Python Code')
         } else {
           $('div.input').show('500');
           $('#toggleButton').val(' Hide Python Code')
```

<IPython.core.display.HTML object>

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```
[3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from IPython.display import display
sns.set_theme()
```

#### 2 Data

The data used in this analysis is retrieved on May 18, 2022 from the website of the Communications Commission of Georgia: comcom.ge.

#### 2.1 Import the data

```
[4]: def import_data_en(excel_path: str) -> pd.DataFrame:
         df = (pd)
               .read_excel(excel_path)
               .sort_values("Month", ascending=True)
               .reset_index(drop=True))
         # English version of the original source contains some
         # untranslated names, fixing it here:
         df = df.replace({
                  ': 'Silknet'.
                    ': 'Globalcell',
                  ': 'Gmobile',
                 ': 'Asanet' # AKA gosim
         })
         # the 'Subscribers' column actually contains subscriber information
         # per company, renaming it accordingly:
         df.rename({'Subscribers': 'CompanySubscribers'}, axis=1, inplace=True)
         return df
```

```
[5]: all_df = import_data_en('data/EN/all.xlsx')
legal_entities_df = import_data_en('data/EN/legal_entities.xlsx')
natural_persons_df = import_data_en('data/EN/natural_persons.xlsx')
all_df.head()
```

```
[5]:
            Month
                        Company
                                 CompanySubscribers
     0 2010-01-01 Veon Georgia
                                             352683
     1 2010-01-01
                       Magticom
                                            1365722
     2 2010-01-01
                        Geocell
                                            1390239
     3 2010-02-01 Veon Georgia
                                             383348
     4 2010-02-01
                       Magticom
                                            1335633
```

## 2.2 Definition of number of subscribers in the context of the given data:

Total Number of subscribers (sim cards), which, during the corresponding period (month, quarter), have used company services at least once (made or received call, sent or received SMS/MMS), used internet or other value added service, or were charged a subscription fee (including company employees and excluding test cards).

#### 2.3 Interval of time covered by the given data

```
[6]: start_month, end_month = all_df['Month'].min(), all_df['Month'].max()

print("\nGiven data covers the interval of time "
    f"from {start_month:%B %Y} to {end_month:%B %Y} (inclusive).\n")
```

Given data covers the interval of time from January 2010 to February 2022 (inclusive).

#### 2.4 Mobile Companies operationg in Georgia

List of companies with their respective number of active months during the time period covered by the given data:

```
Company Active Months
   Veon Georgia
                             146
0
       Magticom
1
                             146
2
        Silknet
                             122
        Geocell
3
                             106
4
     Globalcell
                              86
5
        Gmobile
                              41
         Asanet
```

```
[8]: company_name_list = all_df['Company'].value_counts().index.to_list()

company_colors = {
    'Veon Georgia': 'y',
    'Magticom': 'r',
    'Silknet': 'tab:blue',
    'Geocell': 'tab:purple',
    'Globalcell': 'tab:orange',
    'Gmobile': 'tab:brown',
    'Asanet': 'tab:green',
}
```

Attempting to make colors of companies similar to their respective brand/logo colors:

```
[('Veon Georgia', 'y'),
  ('Magticom', 'r'),
  ('Silknet', 'tab:blue'),
  ('Geocell', 'tab:purple'),
  ('Globalcell', 'tab:orange'),
  ('Gmobile', 'tab:brown'),
  ('Asanet', 'tab:green')]
```

#### 2.5 Date of merger of the Silknet and Geocell

```
[9]: Geocell last month = (all df
                            .loc[all_df['Company'] == 'Geocell', 'Month']
                            .tail(1)
                            .item())
     Silknet_Geocell_merger_month = (all_df
                                      .loc[all_df['Month']>Geocell_last_month,
                                           'Month']
                                      .head(1)
                                      .item())
     def get_Silknet_before merger_subscribers(df: pd.DataFrame) -> int:
         return df.loc[
             (df['Month'] == Geocell last month) & (df['Company'] == 'Silknet'),
             'CompanySubscribers'].item()
     def get_Silknet_Geocell_merger_subscribers(df: pd.DataFrame) -> int:
         return df.loc[(df['Month']==Silknet_Geocell_merger_month)
                       & (df['Company'] == 'Silknet'), 'CompanySubscribers'].item()
     print(f'\n{Silknet_Geocell_merger_month:%B %Y} is the date of Merger'
           ' of the Silknet and Geocell.\n')
```

November 2018 is the date of Merger of the Silknet and Geocell.

#### 2.6 Date of COVID-19 Start

```
[10]: # December 12, 2019:
      # A cluster of patients in Wuhan, Hubei Providence, China
      # begin to experience shortness of breath and fever.
      COVID19_start_date = pd.Timestamp(year=2019, month=12, day=12)
      # February 26, 2020:
      # Georgia officially confirmed its first COVID-19 case.
      # Citizen of Georgia who returned from Iran came back
      # to the Georgian border via Azerbaijan.
      COVID19_start_date_Georgia = pd.Timestamp(year=2020, month=2, day=26)
      def get_total_subscribers_on_the_date(df: pd.DataFrame,
                                            date: pd.Timestamp) -> int:
          # FIXME ignoring edge cases:
          observed_date_before = df.loc[(df['Month'] <= date),</pre>
                                         'Month'].tail(1).item()
          observed_date_after = df.loc[(df['Month'] >= date),
                                        'Month'].head(1).item()
          subscribers_before = df.loc[(df['Month'] == observed_date_before),
                                       'CompanySubscribers'].sum()
          subscribers_after = df.loc[(df['Month'] == observed_date_after),
                                      'CompanySubscribers'].sum()
          return int((subscribers_before+subscribers_after) / 2)
      print(f"COVID-19 start dates:\n - {COVID19_start_date:\%B %d, %Y}: China")
      print(f" - {COVID19_start_date_Georgia: %B %d, %Y}: Georiga")
     COVID-19 start dates:
```

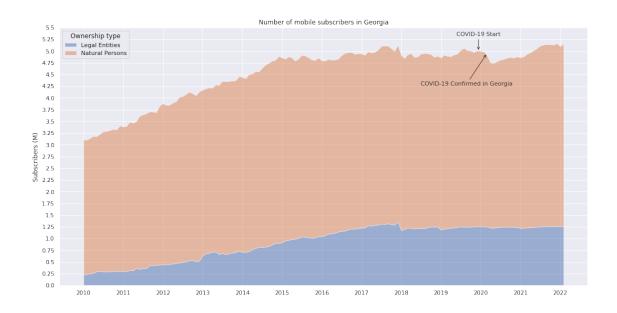
- December 12, 2019: China
- February 26, 2020: Georiga

#### Total number of subscribers

```
[11]: years_df = (all_df
                  .assign(Year=lambda df: [x.year for x in df['Month']])
                  .groupby('Year').first().reset_index()
                  .assign(MonthNum=lambda df: df['Month'].astype(int)))
```

```
[12]: fig = plt.figure(figsize=(18, 9))
      plt.title("Number of mobile subscribers in Georgia")
      plt.stackplot(
          all_df['Month'].unique(),
          legal_entities_df.groupby("Month").sum()['CompanySubscribers'],
          natural_persons_df.groupby("Month").sum()['CompanySubscribers'],
```

```
labels=['Legal Entities', 'Natural Persons'],
   alpha=0.5
COVID19_xy = np.array([
   COVID19_start_date,
   get_total_subscribers_on_the_date(all_df, COVID19_start_date)
])
COVID19_Georgia_xy = np.array([
   COVID19 start date Georgia,
   get_total_subscribers_on_the_date(all_df, COVID19_start_date_Georgia)
])
plt.gca().annotate(
   "COVID-19 Start",
   xy=COVID19_xy,
   xytext=COVID19_xy + np.array([pd.Timedelta(days=0), 3.5e5]),
   arrowprops={'arrowstyle': '->', 'fc': 'k', 'ec': 'k', 'alpha': 0.85},
   fontsize=11, va="center", ha="center", alpha=0.85
plt.gca().annotate(
    "COVID-19 Confirmed in Georgia",
   xy=COVID19_Georgia_xy,
   xytext=COVID19_Georgia_xy - np.array([pd.Timedelta(days=185), 6.5e5]),
   arrowprops={'arrowstyle': '->', 'fc': 'k', 'ec': 'k', 'alpha': 0.85},
   fontsize=11, va="center", ha="center", alpha=0.85
plt.gca().set(
   yticks=np.arange(0, 5.6, 0.25) * 1e6,
   yticklabels=np.arange(0, 5.6, 0.25), ylabel='Subscribers (M)',
   xticks=years_df['Month'], xticklabels=years_df['Year'], xlabel='')
plt.legend(loc='upper left', title='Ownership type')
plt.show()
plt.close(fig)
```

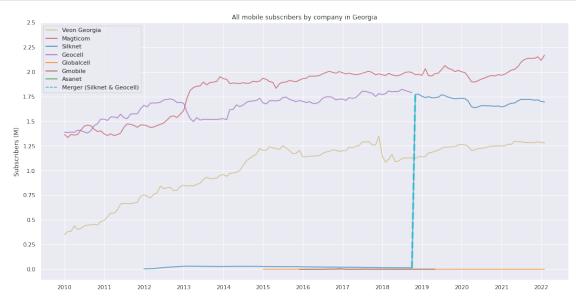


# 4 Number of subscribers by company

#### 4.1 All subscribers

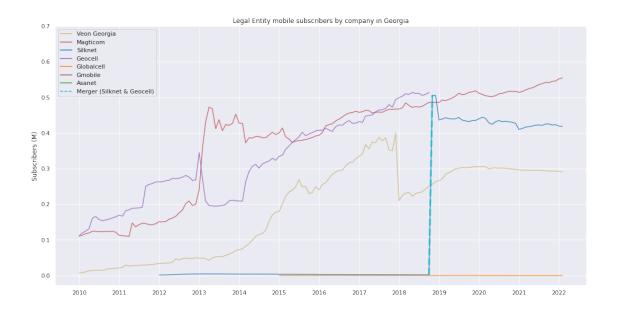
```
[13]: def plot_number_of_subscribers_by_company(
              df: pd.DataFrame(),
              title: str,
              yticks_range: np.ndarray = np.arange(0, 2.75, 0.25)):
          companies = [
              for company in company_name_list if company in df['Company'].unique()]
          company_color_palette = [
              company_colors[company]
              for company in company_name_list if company in companies]
          fig=plt.figure(figsize=(18, 9))
          plt.title(title)
          ax = sns.lineplot(
              data=df,
              x='Month',
              y='CompanySubscribers',
              hue='Company',
              hue_order=companies,
              palette=company_color_palette,
              alpha=0.8,
          )
          with plt.rc_context({'lines.linestyle': '--'}):
              sns.lineplot(
                  x=[Geocell_last_month, Silknet_Geocell_merger_month],
```

```
y=[get_Silknet_before_merger_subscribers(df),
               get_Silknet_Geocell_merger_subscribers(df)],
            hue=['Merger (Silknet & Geocell)', 'Merger (Silknet & Geocell)'],
            palette=['tab:cyan'],
            alpha=0.8,
            linewidth=3.5,
            ax=ax
        )
    ax.set(
        yticks=yticks_range * 1e6,
        yticklabels=np.round(yticks_range, decimals=2),
        ylabel='Subscribers (M)',
        xticks=years_df['Month'], xticklabels=years_df['Year'], xlabel='')
    plt.legend(loc='upper left')
    plt.show()
    plt.close(fig)
plot_number_of_subscribers_by_company(
    all_df,
    "All mobile subscribers by company in Georgia")
```

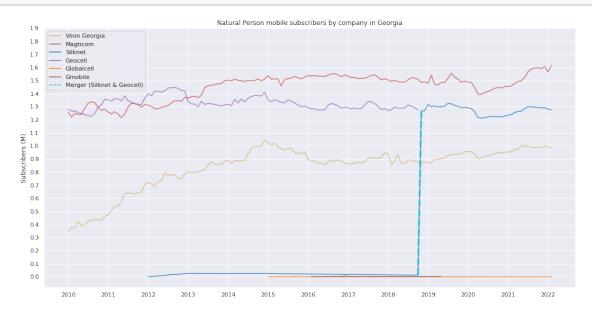


#### 4.2 Legal Entity subscribers

```
[14]: plot_number_of_subscribers_by_company(
    legal_entities_df,
    "Legal Entity mobile subscribers by company in Georgia",
    yticks_range=np.arange(0, 0.8, 0.1))
```



## 4.3 Natural Person subscribers

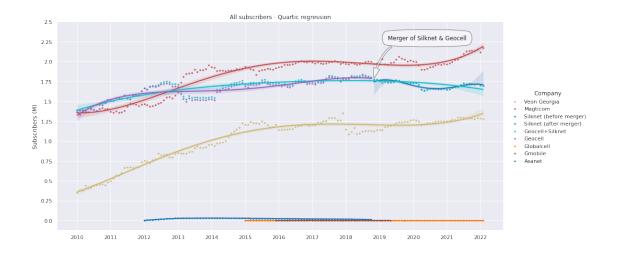


# 5 Trend of number of sybscribers by company

#### 5.1 All subscribers trend

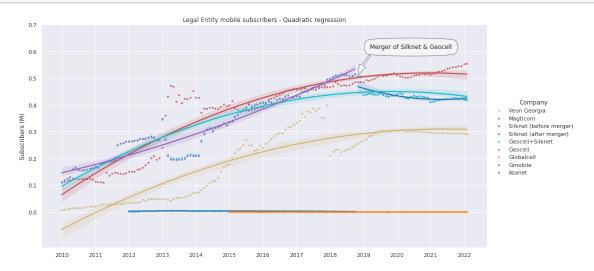
```
[16]: def plot_trend_by_company_regression(
              subscribers_df: pd.DataFrame,
              title: str,
              yticks_range: np.ndarray = np.arange(0, 2.75, 0.25),
              merger_v_offset=5.2e5,
              regression_order=4):
          companies = [
              company
              for company in company_name_list
              if company in subscribers_df['Company'].unique()]
          company_color_palette=[
              company_colors[company]
              for company in company_name_list if company in companies]
          silknet idx = companies.index('Silknet')
          companies[silknet_idx:silknet_idx+1] = [
              'Silknet (before merger)', 'Silknet (after merger)', 'Geocell+Silknet']
          company_color_palette[silknet_idx:silknet_idx+1] = [
              'tab:blue', 'tab:blue', 'tab:cyan']
          df = pd.concat([
              subscribers df,
              subscribers_df
                  .loc[(subscribers_df['Company']=='Silknet')
                       | (subscribers_df['Company'] == 'Geocell')]
                  .groupby('Month').sum().reset_index()
                  .assign(Company='Geocell+Silknet'),
          ], axis=0, ignore_index=True)
          df.loc[(df['Company']=='Silknet')
                 & (df['Month'] < Silknet_Geocell_merger_month),
                 'Company'] = 'Silknet (before merger)'
          df.loc[(df['Company']=='Silknet'), 'Company'] = 'Silknet (after merger)'
          df = df.assign(MonthNum=lambda df: df['Month'].astype(int))
          grid = sns.lmplot(
              data=df,
              x='MonthNum',
              y='CompanySubscribers',
              hue='Company',
              hue_order=companies,
              palette=company_color_palette,
              height=7, aspect=2,
              ci=99,
              scatter_kws={"s": 8},
              order=regression_order,
```

```
plt.title(title + " - {} regression".format(
        {
            1: 'Linear',
            2: 'Quadratic',
            3: 'Cubic',
            4: 'Quartic',
            5: 'Quintic',
            6: 'Sextic',
        }[regression_order]))
    grid.ax.set(
        yticks=yticks_range * 1e6,
        yticklabels=np.round(yticks_range, decimals=2),
        ylabel='Subscribers (M)',
        xticks=years_df['MonthNum'], xticklabels=years_df['Year'], xlabel='')
    merger_xy = np.array([
        df.loc[df['Month'] == Silknet_Geocell_merger_month,
               'MonthNum'].head(1).item(),
        get_Silknet_Geocell_merger_subscribers(subscribers_df)])
    grid.ax.annotate(
        "Merger of Silknet & Geocell",
        xy=merger_xy,
        xytext=merger_xy + np.array([5e16, merger_v_offset]),
        arrowprops={
            'arrowstyle': ('simple,head_length=1.8,'
                           'head_width=1.2,tail_width=0.3'),
            'connectionstyle': 'arc3,rad=0.45',
            'fc': 'w', 'ec': 'k', 'lw': 1.1, 'alpha': 0.5,
        },
        bbox={'boxstyle': 'round4', 'alpha': 0.5, 'pad': 1.1,
              'fc': 'w', 'ec': 'k', 'lw': 1.1},
        fontsize=12, va="center", ha="center",
    plt.show()
    plt.close()
plot_trend_by_company_regression(all_df, "All subscribers", regression_order=4)
```



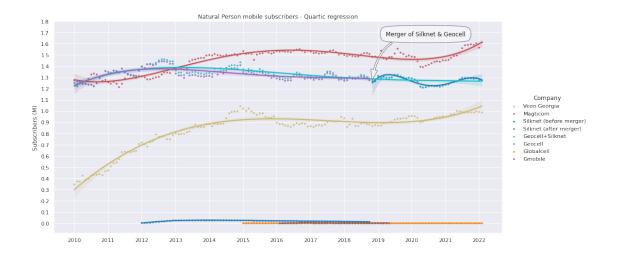
## 5.2 Legal Entity subscribers trend

```
[17]: plot_trend_by_company_regression(
    legal_entities_df, "Legal Entity mobile subscribers",
    yticks_range=np.arange(0, 0.8, 0.1), merger_v_offset=1.1e5,
    regression_order=2)
```



#### 5.3 Natural Person subscribers trend

```
[18]: plot_trend_by_company_regression(
    natural_persons_df, "Natural Person mobile subscribers",
    yticks_range=np.arange(0, 1.9, 0.1), merger_v_offset=4.2e5,
    regression_order=4)
```



# 6 Performance of Companies

```
[19]: def get_total_subscribers_df(df: pd.DataFrame) -> pd.DataFrame:
          return (df
                  .groupby('Month')['CompanySubscribers']
                  .sum()
                  .reset index()
                  .rename({'CompanySubscribers': 'MarketSubscribers'}, axis=1)
                  .assign(
                      MarketSubscribersDelta=lambda df: np.concatenate([
                          np.array([0]),
                          df.iloc[1:]['MarketSubscribers'].to_numpy()
                          - df.iloc[:-1]['MarketSubscribers'].to_numpy()
                      ])
                  )
                 )
      def compute_subscriber_and_market_features(df: pd.DataFrame) -> pd.DataFrame:
          df = (df)
                .join(get_total_subscribers_df(df)[['Month', 'MarketSubscribers',
                                                     'MarketSubscribersDelta'll
                      .set_index('Month'), on='Month')
                .groupby('Company')
                .apply(lambda group_df: group_df.assign(
                    CompanySubscribersDelta=lambda df: np.concatenate([
                        np.array([0]),
                        (df.iloc[1:]['CompanySubscribers'].to numpy()
                         - df.iloc[:-1]['CompanySubscribers'].to_numpy())
                    ])
```

```
.assign(
              CompanySubscribersShare=lambda df: (df['CompanySubscribers']
                                                   / df['MarketSubscribers']),
              CompanySubscribersShareOld=lambda df: np.concatenate([
                  np.array([0]),
                  df['CompanySubscribersShare'].to_numpy()[:-1]
              ])
          )
         )
    df.loc[(df['Company']=='Silknet')
           & (df['Month'] == Silknet_Geocell_merger_month),
           'CompanySubscribersDelta'] = 0 # merger... not an actual growth...
    return df
print("New/Changed columns after applying the "
      "`compute_subscriber_and_market_features()` function to DataFrame:\n-",
      "\n- ".join(list(compute subscriber and market features(all df).columns
                       .difference(all_df.columns))))
```

New/Changed columns after applying the

`compute\_subscriber\_and\_market\_features()` function to DataFrame:

- CompanySubscribersDelta
- CompanySubscribersShare
- CompanySubscribersShareOld
- MarketSubscribers
- MarketSubscribersDelta

Descriptive statistics of monthly changes in the number of All Subscribers:

	MarketSubscribersDelta	CompanySubscribersDelta
count	648.00	648.00
mean	11789.52	3189.55
std	50117.14	21336.14
min	-221244.00	-202065.00
25%	-14333.00	-1159.50
50%	13623.00	13.00
75%	44137.00	8334.75
max	136533.00	128717.00

Descriptive statistics of monthly changes in the number of Legal Entity Subscribers:

	${\tt MarketSubscribersDelta}$	${\tt CompanySubscribersDelta}$
count	608.00	608.00
mean	7090.89	1716.18
std	26092.10	13841.48
min	-184415.00	-188934.00
25%	-1151.00	-72.00
50%	6477.50	108.50
75%	16116.00	3833.00
max	115523.00	118730.00

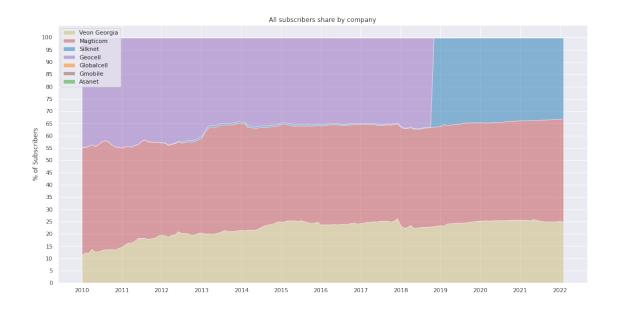
Descriptive statistics of monthly changes in the number of Natural Person Subscribers:

	MarketSubscribersDelta	CompanySubscribersDelta
count	646.00	646.00
mean	4840.02	1584.16
std	41158.06	16627.51
min	-132445.00	-84790.00
25%	-23975.50	-3381.00
50%	4341.00	-1.00
75%	32419.00	5354.75
max	127319.00	61788.00

# 7 Share Of Subscribers by company

#### 7.1 All subscribers share

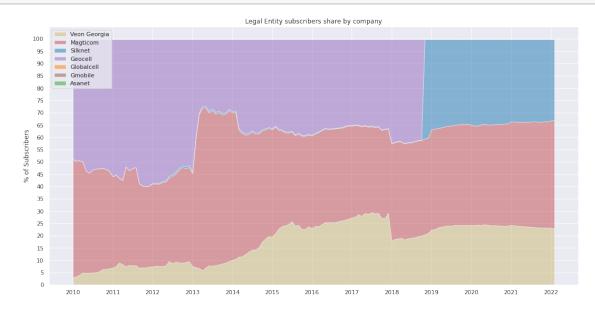
```
[23]: def plot_market_share_of_companies(df: pd.DataFrame, title: str):
          df = compute_subscriber_and_market_features(df)
          companies = [
              company for company in company_name_list
              if company in df['Company'].unique()]
          company_color_palette=[
              company_colors[company] for company in company_name_list
              if company in companies]
          company_subscribers = (df[['Month','Company','CompanySubscribersShare']]
                                 .pivot(index='Month',columns='Company')
                                 .T.reset index(drop=True, level=0).T
                                 .fillna(0)[companies]
                                 .to_dict(orient='list')
          fig=plt.figure(figsize=(18,9))
          plt.title(title)
          plt.stackplot(
              df['Month'].unique(),
              company_subscribers.values(),
              labels=company_subscribers.keys(),
              colors=company_color_palette,
              alpha=0.5
          )
          plt.gca().set(
              yticks=np.arange(0, 1.01, 0.05), yticklabels=np.arange(0, 101, 5),
              ylabel='% of Subscribers',
              xticks=years_df['Month'], xticklabels=years_df['Year'], xlabel='')
          plt.legend(loc='upper left')
          plt.show()
          plt.close(fig)
      plot_market_share_of_companies(all_df, "All subscribers share by company")
```



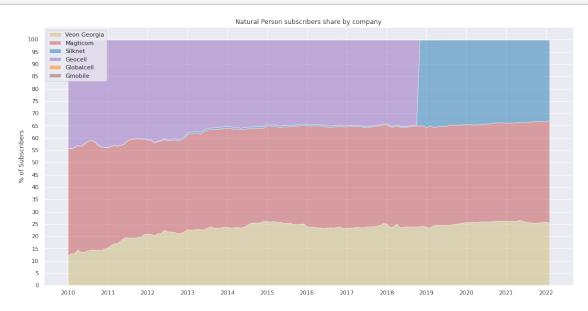
# 7.2 Legal Entity subscribers share

[24]: plot\_market\_share\_of\_companies(legal\_entities\_df,

"Legal Entity subscribers share by company")



#### 7.3 Natural Person subscribers share



#### 8 The formula of The Atractiveness Score

NOTE: This is a very simple and straightforward formula, so I did not attempt to research if this or variations of it already exist, I just created it independently. If the formula is familiar to you from any other credible sources, please provide your feedback by the means indicated in the end of this notebook in order for me to give a due credit to whoever might be using the exact same formula earlier.

"Attractiveness Score" as used in this notebook is intended to be a company performance metric measuring an ability of a company to literally attract subscribers and further retain them, generating a growth of the number of its subscribers in the process.

First, let's define an "Attractiveness Score" of a company in terms of the variable/feature names available and computed in this notebook:

AttractivenessScore = CompanySubscribersDelta - (0.5 \* ('CompanySubscribersShareOld' + 'CompanySubscribersShare') \* MarketSubscribersDelta)

Where the CompanySubscribersDelta is a change in number of subscribers of a company in a given period (Month/Year), the MarketSubscribersDelta is a change in number of subscribers in the total market in the same period and the 0.5 \* ('CompanySubscribersShareOld' + 'CompanySubscribersShare') is an average share of subscribers of a company out of the total market in the same period.

Second, let's define mathematically the same formula for the sake of clarity:

Let  $n^c$  be a number of subscribers of a company and let  $n^m$  be a number of total subscribers in the market (i.e. sum of subscribers of all companies in the market). Then we can define the AtractivenessScore of a company for any time period T,  $AtractivenessScore_T$ , as follows:

$$AtractivenessScore_T = \sum_{i=1}^{N} \big(n^c_{t_i} - n^c_{t_{i-1}}\big) - \frac{1}{2} \big(\frac{n^c_{t_{i-1}}}{n^m_{t_{i-1}}} + \frac{n^c_{t_i}}{n^m_{t_i}}\big) \big(n^m_{t_i} - n^m_{t_{i-1}}\big)$$

where T is a given time period (Month/Quarter, Year etc.) for which the score is calculated;  $N>0,\ N\in\mathbb{N}$  is a number of observations made within the T time period;  $\{t_1,t_2,t_3,t_N\}\in T$ are respective times of each observation during the time period;  $i > j \iff t_i > t_j$  i.e. observation times are indexed from oldest to newest;  $n^{c_{t_i}} \in \mathbb{N}, n^{c_{t_i}} \geq 0$  is an observed number of subscribers of a company for the respective  $t_i$  time;  $n_{t_i}^m \in \mathbb{N}, \ n_{t_i}^m \geq 0$  is an observed total number of subscribers in the market for the respective  $t_i$  time;

By denoting  $\Delta n^c_{t_i} = (n^c_{t_i} - n^c_{t_{i-1}}), \ \Delta n^m_{t_i} = (n^m_{t_i} - n^m_{t_{i-1}})$  and  $\bar{s}^c_{t_i} = \frac{1}{2}(\frac{n^c_{t_{i-1}}}{n^m_{t_{i-1}}} + \frac{n^c_{t_i}}{n^m_{t_i}})$  the above formula becomes:

$$At ractiveness Score_T = \sum_{i=1}^{N} \Delta n_{t_i}^c - \bar{s}_{t_i}^c \Delta n_{t_i}^m$$

Finally, going back from the mathematical formula to the variable/feature names actually used in this notebook, the correspondence between the two is as follows:

- $\begin{array}{l} \bullet \ \, \text{CompanySubscribersDelta} \iff \Delta n_{t_i}^c \\ \bullet \ \, \text{MarketSubscribersDelta} \iff \Delta n_{t_i}^m \\ \bullet \ \, \text{0.5 * ('CompanySubscribersShareOld' + 'CompanySubscribersShare')} \iff \bar{s}_{t_i}^c \\ \end{array}$

I added the  $\Delta n_t^c$  component to the formula for an obvious reasons, as it is a change in number of subscribers of a company, which should be by definition one of the best indicators of attractiveness of a company, but it can contain noises, such as noises originated from growing or shrinking of the market itself (i.e. change in total number of subscribers in the market), so I added the  $-\bar{s}_t^c \Delta n_t^m$ component of the formula with the purpose to absorb some of those noises and yield more accurate metric better focusing on the portion of a change in number of subscribers which is actually driven by the attractiveness of the company itself and not by external factors like shrinking or groving market. Assumption behind this part of the formula is that in the events of growing or shrinking of the market if the companies are equally attractive then they would grow or shrink proportionally.

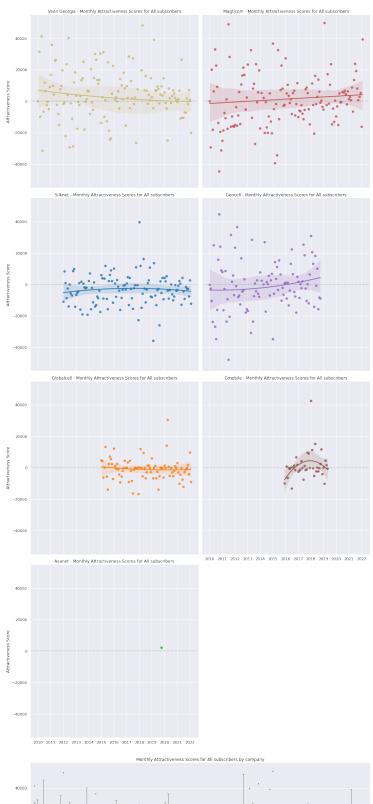
This formula is influenced by the data given in this analysis, as it was made speciphically for this data, which contains a limited information (i.e. mainly a number of subscribers), the intention was to make formula able to extract valuable insights from the given data. Although we cannot access directly performances of companies in the  $PR \ \mathcal{E}$  marketing or customer satisfaction and other areas of their operations just by using this limited data, we still can use the AttractivenessScore obtained with this formula as a proxy into those performances combined without need for a more information, and as the formula also cancels some inherent noise of the data, the resulting metric is supposedly more close to a real world performance of the scored companies in terms of attracting subscribers.

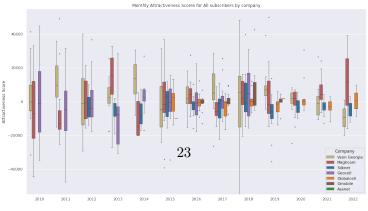
```
[26]:
     def compute_attractiveness_score(df: pd.DataFrame) -> pd.DataFrame:
          return compute_subscriber_and_market_features(df).assign(
              AttractivenessScore = lambda df: (
                  df['CompanySubscribersDelta']
```

## 9 Monthly Attractiveness Scores of Companies

#### 9.1 Monthly Scores for All subscribers

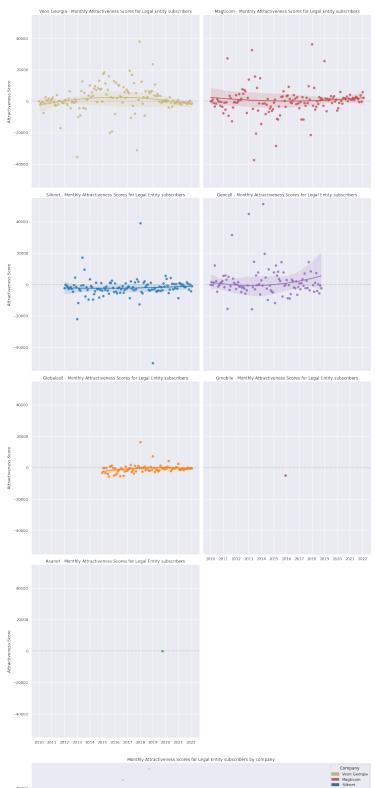
```
[27]: display(compute attractiveness_score(all_df)['AttractivenessScore'].describe())
     count
                 648.000000
                  90.636727
     mean
               18255.137182
     std
             -176201.147065
     min
     25%
               -7301.075624
     50%
                 -20.258358
     75%
                6489.583900
     max
              120434.423264
     Name: AttractivenessScore, dtype: float64
[28]: def plot_attractiveness_score_of_companies(df: pd.DataFrame, title: str):
          df = df.assign(MonthNum=lambda df: df['Month'].astype(int),
                         Year=lambda df: [x.year for x in df['Month']])
          df = compute_attractiveness_score(df)
          companies = [
              company for company in company_name_list
              if company in df['Company'].unique()]
          company color palette = [
              company_colors[company] for company in company_name_list
              if company in companies]
          grid = sns.FacetGrid(df, col="Company", hue="Company", col_order=companies,
                               hue_order=companies, palette=company_color_palette,
                               col_wrap=2, height=7.5)
          grid.refline(y=0, linestyle=":") # Draw a horizontal line at zero
          grid.map(sns.regplot, 'MonthNum', 'AttractivenessScore', marker='o',
                   order=2, ci=98)
          grid.set_titles(col_template=f"{{col_name}} - {title}",
                          row_template="{row_name}")
          grid.set(
              ylabel='Attractiveness Score',
              xticks=years_df['MonthNum'], xticklabels=years_df['Year'], xlabel='',
              ylim=(-55000, 55000)
          )
```

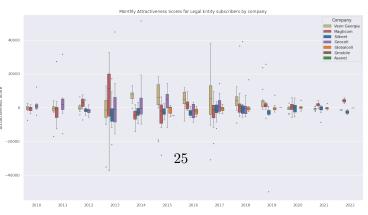




## 9.2 Monthly Scores for Legal Entity subscribers

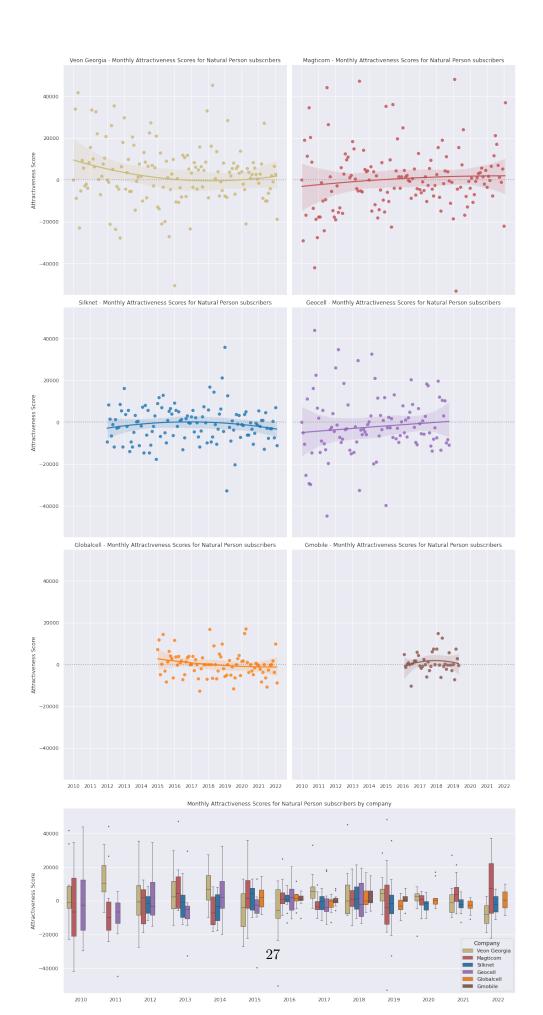
```
[29]: display(compute_attractiveness_score(
          legal_entities_df)['AttractivenessScore'].describe())
     count
                 608.000000
     mean
                 122.907280
     std
               12269.675349
             -135937.797389
     min
     25%
               -2536.133414
     50%
                -262.757392
     75%
                2240.071951
              105826.940724
     max
     Name: AttractivenessScore, dtype: float64
[30]: plot_attractiveness_score_of_companies(
          legal_entities_df,
          "Monthly Attractiveness Scores for Legal Entity subscribers")
```





## 9.3 Monthly Scores for Natural Person subscribers

```
[31]: display(compute_attractiveness_score(
          natural_persons_df)['AttractivenessScore'].describe())
                646.000000
     count
     mean
                 57.866318
              13140.832671
     std
             -70452.871477
     min
     25%
              -6092.321398
     50%
                 -1.850429
     75%
               5634.597747
              48164.333101
     max
     Name: AttractivenessScore, dtype: float64
[32]: plot_attractiveness_score_of_companies(
          natural_persons_df,
          "Monthly Attractiveness Scores for Natural Person subscribers")
```



## 10 Yearly Attractiveness Scores of Companies

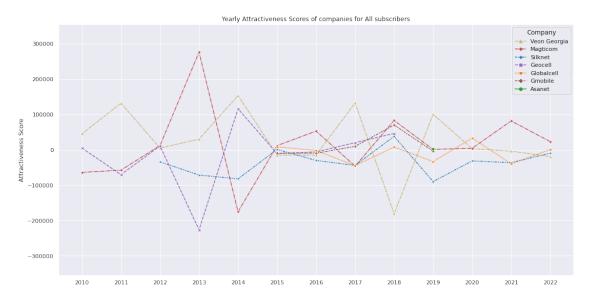
#### 10.1 Yearly Scores for All subscribers

```
[33]: def get_top_3_companies_per_year(df: pd.DataFrame) -> pd.DataFrame:
          df = df.assign(
              MonthNum=lambda df: df['Month'].astype(int),
              Year=lambda df: [x.year for x in df['Month']],
          )
          df = (compute_attractiveness_score(df)
                .groupby(['Company', 'Year'])
                .aggregate({'AttractivenessScore': 'sum'}).reset_index()
                .groupby('Year')
                .apply(lambda df: (df
                                    .sort_values('AttractivenessScore',
                                                 ascending=False)
                                    .head(3)
                                    .reset index(drop=True)))
                .reset_index(level=1).rename({'level_1': 'Place'}, axis=1)
                .pivot('Year', 'Place').rename({0: 1, 1: 2, 2: 3}, axis=1)
          )
          return df
      def plot_yearly_attractiveness score_of_companies(df: pd.DataFrame,
                                                         title: str):
          df = df.assign(
              MonthNum=lambda df: df['Month'].astype(int),
              Year=lambda df: [x.year for x in df['Month']]
          )
          df = compute_attractiveness_score(df)
          df = (df.groupby(['Company', 'Year'])
                .aggregate({'AttractivenessScore': 'sum'}).reset index())
          companies = [
              company for company in company_name_list
              if company in df['Company'].unique()]
          company_color_palette = [
              company_colors[company] for company in company_name_list
              if company in companies]
          fig = plt.figure(figsize=(18, 9))
          ax = sns.lineplot(data=df, x='Year', y='AttractivenessScore',
```

Yearly Attractiveness Scores of companies for All subscribers:

	Company			AttractivenessScore	\
Place	1	2	3	1	
Year					
2010	Veon Georgia	Geocell	Magticom	45362.083982	
2011	Veon Georgia	Magticom	Geocell	131984.657940	
2012	Magticom	Geocell	Veon Georgia	12136.594211	
2013	Magticom	Veon Georgia	Silknet	276248.323637	
2014	Veon Georgia	Geocell	Silknet	153741.442115	
2015	Magticom	Globalcell	Silknet	11834.482137	
2016	Magticom	Globalcell	Geocell	52927.240468	
2017	Veon Georgia	Geocell	Gmobile	132636.468918	
2018	Magticom	Gmobile	Geocell	82995.183958	
2019	Veon Georgia	Asanet	Magticom	100520.205002	
2020	Globalcell	${ t Magticom}$	Veon Georgia	33395.204569	
2021	Magticom	Veon Georgia	Silknet	81731.975640	
2022	Magticom	Globalcell	Silknet	23383.519692	
Place	2	3			
Year					
2010	4856.801673	-63840.835359			
2011	-57561.318275	-70160.992134			
2012	10839.125704	4863.307230			
2013	30273.915515	-71402.706360			
2014	115931.736293	-82111.949335			
2015	8480.221206	1293.094662			

```
2016
        -1778.601784 -5913.157603
2017
        20202.770120
                       9713.357893
2018
        70269.224326 46335.943660
2019
         2197.154207
                       1197.475701
         4339.903478
                       3196.041334
2020
2021
        -3725.224280 -36470.176832
2022
          854.293360
                      -9328.489339
```



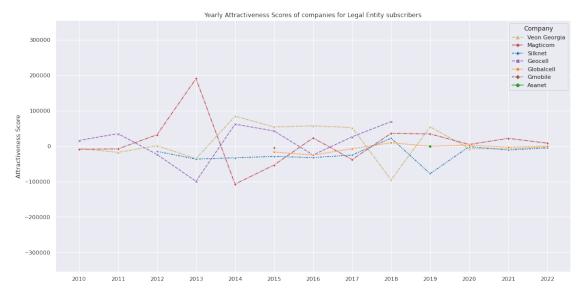
#### 10.2 Yearly Scores for Legal Entity subscribers

Yearly Attractiveness Scores of companies for Legal Entity subscribers:

	Company			${\tt AttractivenessScore}$	\
Place	1	2	3	1	
Year					
2010	Geocell	Veon Georgia	Magticom	16015.083230	
2011	Geocell	${ t Magticom}$	Veon Georgia	35038.238082	
2012	Magticom	Veon Georgia	Silknet	32021.290422	
2013	Magticom	Veon Georgia	Silknet	190397.530087	
2014	Veon Georgia	Geocell	Silknet	84478.617926	
2015	Veon Georgia	Geocell	Gmobile	54016.232382	
2016	Veon Georgia	Magticom	Geocell	57131.393478	

2017	Veon Georgia	Geocell	Globalcell	52236.986105
2018	Geocell	${ t Magticom}$	Silknet	69153.218138
2019	Veon Georgia	${ t Magticom}$	Asanet	54390.024447
2020	Magticom	Globalcell	Silknet	4486.802501
2021	${ t Magticom}$	Globalcell	Veon Georgia	21836.080569
2022	${ t Magticom}$	Globalcell	Veon Georgia	8245.535916

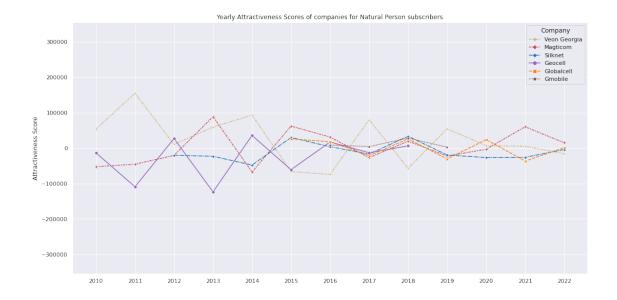
Place	2	3
Year		
2010	-6592.335644	-9024.622081
2011	-8047.902630	-17747.657414
2012	1166.772836	-15008.081363
2013	-35643.258459	-36776.163718
2014	61890.963138	-33149.647736
2015	42566.644955	-4849.277917
2016	22946.183006	-25146.310677
2017	26132.037937	-7363.107595
2018	35778.728272	22243.376309
2019	34327.040916	-0.456368
2020	2923.260004	-1441.009696
2021	-3386.336872	-8133.935355
2022	-247.091267	-2759.755264



## 10.3 Yearly Scores for Natural Person subscribers

Yearly Attractiveness Scores of companies for Natural Person subscribers:

	Company			AttractivenessScore	\
Place	1	2	3	1	
Year					
2010	Geocell	Veon Georgia	Magticom	16015.083230	
2011	Geocell	Magticom	Veon Georgia	35038.238082	
2012	Magticom	Veon Georgia	Silknet	32021.290422	
2013	Magticom	Veon Georgia	Silknet	190397.530087	
2014	Veon Georgia	Geocell	Silknet	84478.617926	
2015	Veon Georgia	Geocell	Gmobile	54016.232382	
2016	Veon Georgia	Magticom	Geocell	57131.393478	
2017	Veon Georgia	Geocell	Globalcell	52236.986105	
2018	Geocell	Magticom	Silknet	69153.218138	
2019	Veon Georgia	Magticom	Asanet	54390.024447	
2020	Magticom	Globalcell	Silknet	4486.802501	
2021	Magticom	Globalcell	Veon Georgia	21836.080569	
2022	Magticom	Globalcell	Veon Georgia	8245.535916	
Place	2	3			
Year					
2010	-6592.335644	-9024.622081			
2011	-8047.902630	_,,,,,,,,,,			
2012		-15008.081363			
2013		-36776.163718			
2014		-33149.647736			
2015	42566.644955	-4849.277917			
2016	22946.183006	-25146.310677			
2017	26132.037937	-7363.107595			
2018	35778.728272	22243.376309			
2019	34327.040916	-0.456368			
2020	2923.260004	-1441.009696			
2021	-3386.336872	-8133.935355			
2022	-247.091267	-2759.755264			



# 11 Top Companies by Attractiveness Score

```
[36]: def get_top_companies_per_year(df: pd.DataFrame, name: str) -> pd.DataFrame:
          return (get_top_3_companies_per_year(df)
                  .loc[:,[('Company',1)]].T.reset_index(drop=True).T
                  .rename({0:name}, axis=1))
      def extract_available_months(df: pd.DataFrame) -> pd.DataFrame:
          return (df
                  .assign(
                      MonthName=lambda df: [x.month_name() for x in df['Month']],
                      Year=lambda df: [x.year for x in df['Month']],
                  .groupby('Year')
                  .apply(lambda df: pd.DataFrame({
                      "Months": [
                           ' - '.join((lambda m: [m[0], m[-1]] if len(m) > 1 else m)(
                              df['MonthName'].unique()
                          ))
                      ]
                  })).reset_index()
                 ).set_index('Year').filter(['Months'])
      print("\nTop Mobile Communication Companies in Georgia "
            "by Attractiveness Score per year:\n")
      top_companies_df = get_top_companies_per_year(all_df,
```

```
"For All subscribers").join(
    get_top_companies_per_year(legal_entities_df,
                               "For Legal Entity subscribers")
).join(
    get_top_companies_per_year(natural_persons_df,
                               "For Natural Person subscribers")
).join(
    extract_available_months(all_df)
).reset_index().set_index(['Months', 'Year'])
display(top_companies_df)
for category in ('For All subscribers', 'For Legal Entity subscribers',
                 'For Natural Person subscribers'):
    print(f'{category} during the '
          f'{top_companies_df.loc["January - December"].index.min()}'
          f' - {top_companies_df.loc["January - December"].index.max()} years '
          'the following companies held the yearly top spot:')
    for company,count in top_companies_df.loc['January - December',
                                               category
                                             ].value_counts().iteritems():
        company = '"{}"'.format(company)
        print(f' - {company:<16} {count:} {"times" if count > 1 else "time"}.')
    print()
```

Top Mobile Communication Companies in Georgia by Attractiveness Score per year:

```
For All subscribers For Legal Entity subscribers
Months
                    Year
January - December 2010
                                Veon Georgia
                                                                    Geocell
                    2011
                                Veon Georgia
                                                                    Geocell
                    2012
                                    Magticom
                                                                   Magticom
                   2013
                                    Magticom
                                                                   Magticom
                                Veon Georgia
                                                              Veon Georgia
                   2014
                    2015
                                    Magticom
                                                              Veon Georgia
                   2016
                                    Magticom
                                                              Veon Georgia
                    2017
                                Veon Georgia
                                                              Veon Georgia
                   2018
                                    Magticom
                                                                    Geocell
                    2019
                                Veon Georgia
                                                              Veon Georgia
                    2020
                                  Globalcell
                                                                   Magticom
                    2021
                                    Magticom
                                                                   Magticom
January - February 2022
                                                                   Magticom
                                    Magticom
                         For Natural Person subscribers
Months
                    Year
January - December 2010
                                            Veon Georgia
```

```
2012
                                                     Geocell
                         2013
                                                    Magticom
                        2014
                                                Veon Georgia
                        2015
                                                    Magticom
                        2016
                                                    Magticom
                        2017
                                                Veon Georgia
                        2018
                                                     Silknet
                        2019
                                                Veon Georgia
                         2020
                                                  Globalcell
                         2021
                                                    Magticom
     January - February 2022
                                                    Magticom
     For All subscribers during the 2010 - 2021 years the following companies held
     the yearly top spot:
      - "Magticom"
                         6 times.
      - "Veon Georgia"
                         5 times.
      - "Globalcell"
                         1 time.
     For Legal Entity subscribers during the 2010 - 2021 years the following
     companies held the yearly top spot:
      - "Veon Georgia" 5 times.
      - "Magticom"
                         4 times.
      - "Geocell"
                         3 times.
     For Natural Person subscribers during the 2010 - 2021 years the following
     companies held the yearly top spot:
      - "Veon Georgia"
                         5 times.
      - "Magticom"
                         4 times.
      - "Geocell"
                         1 time.
      - "Silknet"
                         1 time.
      - "Globalcell"
                         1 time.
[37]: print(f"\n ** Total Elapsed time: {datetime.utcnow() - nb st} ** \n")
      print(f"Notebook END time: {datetime.utcnow()} UTC\n")
      ** Total Elapsed time: 0:00:16.340084 **
     Notebook END time: 2022-06-30 01:43:47.510279 UTC
[38]: %%capture
      %mkdir OGP_classic
[39]: %%capture
      %%file "OGP_classic/conf.json"
```

Veon Georgia

2011

```
{
  "base_template": "classic",
  "preprocessors": {
     "500-metadata": {
        "type": "nbconvert.preprocessors.ClearMetadataPreprocessor",
        "enabled": true,
        "clear_notebook_metadata": true,
        "clear_cell_metadata": true
    },
     "900-files": {
        "type": "nbconvert.preprocessors.ExtractOutputPreprocessor",
        "enabled": true
    }
}
```

```
[40]: %%capture
      %%file "OGP_classic/index.html.j2"
      {%- extends 'classic/index.html.j2' -%}
      {%- block html_head -%}
      {# OGP attributes for shareability #}
      <meta property="og:url"</pre>
                                         content="https://sentinel-1.github.io/
       →mobile subscribers Georgia/" />
      <meta property="og:type"</pre>
                                         content="article" />
      <meta property="og:title"</pre>
                                         content="Ranking of Mobile Operators Companies_
       ⇔in Georgia" />
      <meta property="og:description" content="Which companies are preffered by__</pre>
       →mobile subscribers in Georgia?" />
      <meta property="og:image"</pre>
                                         content="https://raw.githubusercontent.com/
       →sentinel-1/mobile_subscribers_Georgia/master/images/ASA_Mobile_123.png" />
      <meta property="og:image:alt"</pre>
                                         content="ASA Mobile Companies Ranking 123_
       ⇔logo" />
      <meta property="og:image:type"</pre>
                                         content="image/png" />
      <meta property="og:image:width" content="1200" />
      <meta property="og:image:height" content="628" />
      <meta property="article:published_time" content="2022-05-23T22:33:25+00:00" />
      <meta property="article:modified_time" content="{{ resources.</pre>
       →iso8610_datetime_now }}" />
      <meta property="article:publisher"</pre>
                                                content="https://sentinel-1.github.io" /
      <meta property="article:author"</pre>
                                                content="https://github.com/sentinel-1"
       →/>
      <meta property="article:section"</pre>
                                                content="datascience" />
      <meta property="article:tag"</pre>
                                                content="datascience" />
```

```
content="Python" />
<meta property="article:tag"</pre>
                                       content="data" />
<meta property="article:tag"</pre>
<meta property="article:tag"</pre>
                                       content="analytics" />
<meta property="article:tag"</pre>
                                       content="datavisualization" />
<meta property="article:tag"</pre>
                                       content="bigdataunit" />
<meta property="article:tag"</pre>
                                       content="visualization" />
<meta property="article:tag"</pre>
                                       content="mobilesubscribers" />
<meta property="article:tag"</pre>
                                       content="mobilecompanies" />
<meta property="article:tag"</pre>
                                       content="ranking" />
<meta property="article:tag"</pre>
                                       content="telecomunication" />
<link rel="icon" type="image/x-icon" href="../favicon.ico">
{{ super() }}
{\"- endblock html_head -\"\}
{% block body_header %}
<body>
<div class="container">
  <nav class="navbar navbar-default">
   <div class="container-fluid">
      role="presentation">
         <a href="/">
           <svg xmlns="http://www.w3.org/2000/svg"</pre>
                viewBox="0 0 576 512" width="1em">
             <path
               fill="#999999"
\mathtt{d="M}\ 288,0\ 574,288\ 511,288\ 511,511\ 352,511\ 352,352\ 223,352\ 223,511\ 62,511_{\sqcup}
 ⇔64,288 0,288 Z"
             />
           </svg> Home
         </a>
       role="presentation" class="active">
         <a href="/mobile_subscribers_Georgia/"> English </a>
       <a href="/mobile_subscribers_Georgia/ka/">
                                                        </a>
```

```
</div>
  </nav>
</div>
 <div tabindex="-1" id="notebook" class="border-box-sizing">
    <div class="container" id="notebook-container">
{% endblock body_header %}
{% block body_footer %}
   </div>
  </div>
  <footer>
    <div class="container"</pre>
         style="display:flex; flex-direction: row; justify-content: center; _
 →align-items: center;">
       © 2022
        <a href="https://github.com/sentinel-1" target="_blank">Sentinel-1</a>
      <!-- TOP.GE ASYNC COUNTER CODE -->
      <div id="top-ge-counter-container" data-site-id="116052"</pre>
           style="margin-right: 3.7em; float: right; "></div>
      <script async src="//counter.top.ge/counter.js"></script>
      <!-- / END OF TOP.GE COUNTER CODE -->
      <!-- ANALYTICS.LAGOGAL.COM -->
      <div id="analytics-lagogal-com-access" data-site-id="20221"</pre>
           style="margin: 0;padding: 0;"></div>
      <script async src="//analytics.lagogal.com/access.js"></script>
      <!-- / END OF ANALYTICS.LAGOGAL.COM -->
     </div>
  </footer>
</body>
{% endblock body_footer %}
```

This notebook is originally published under the Apache License (Version 2.0) at the following GitHub repository: sentinel-1/mobile\_subscribers\_Georgia

For the issues, feedback or suggestions regarding the original notebook (if any) feel free to open an issue at the corresponding Issues page of the repository