DATA EXPLORATION

Installing dependecies

```
!pip install autocorrect
In [1]:
        !pip install wordcloud
        Requirement already satisfied: autocorrect in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (2.
        1.0)
        Requirement already satisfied: wordcloud in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (1.8.
        Requirement already satisfied: numpy>=1.6.1 in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (f
        rom wordcloud) (1.17.4)
        Requirement already satisfied: matplotlib in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (fro
        m wordcloud) (2.2.2)
        Requirement already satisfied: pillow in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (from wo
        rdcloud) (7.0.0)
        Requirement already satisfied: cycler>=0.10 in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (f
        rom matplotlib->wordcloud) (0.10.0)
        Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in c:\users\henry\anaconda3\envs\ecog
        eo tool\lib\site-packages (from matplotlib->wordcloud) (2.4.6)
        Requirement already satisfied: python-dateutil>=2.1 in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-pac
        kages (from matplotlib->wordcloud) (2.8.1)
        Requirement already satisfied: pytz in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (from matp
        lotlib->wordcloud) (2019.3)
        Requirement already satisfied: six>=1.10 in c:\users\henry\anaconda3\envs\ecogeo_tool\lib\site-packages (from
        matplotlib->wordcloud) (1.13.0)
        Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\henry\anaconda3\envs\ecogeo_tool\lib\site-packag
        es (from matplotlib->wordcloud) (1.1.0)
        Requirement already satisfied: setuptools in c:\users\henry\anaconda3\envs\ecogeo tool\lib\site-packages (fro
        m kiwisolver>=1.0.1->matplotlib->wordcloud) (44.0.0.post20200106)
```

Loading dependecies

```
In [2]: from data_exploring.data_exploring_functions import *
    import plotly.express as px
    import nltk
    from wordcloud import WordCloud
    import os

[nltk_data] Downloading package stopwords to
    [nltk_data] C:\Users\henry\AppData\Roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

Loading and cleaning data

```
In [3]: #File names and base path.
    file_satisfaction = 'satisfaction_ratings'
    file_response = 'NPS_responses'
    base_datapath = os.path.join(os.getcwd(),'data_exploring','data')

#Loading the data using a function from data_exploring_functions. Spell correction, lower letters and elimina
    tion of stopwrods is done in
    #this step. It returns the loaded dataset with an extra column named "processed comment"
    data_satisfaction = loading_data(file_satisfaction,base_datapath,'csv')
    data_responses = loading_data(file_response,base_datapath,'csv')
```

Data visualization

In [4]: data_satisfaction.head(3)

Out[4]:

	Unnamed: 0	Requester	User Id	Email	Ticket Id	Brand	Group	Assignee	Satisfaction	Cc
0	0	Carlos Enrique brito	401415540751	carlosenrique1989n@hotmail.com	595380	OFFCORSS	Soporte OFFCORSS	CTS Transporte	good	
1	1	Lizeth Herrera	401737616791	lkhr328@gmail.com	600778	OFFCORSS	Soporte OFFCORSS	Devoluciones	good	
2	2	Carolina Sanchez	401569843672	caritoss1@hotmail.com	598181	OFFCORSS	Soporte OFFCORSS	CTS Transporte	bad	ех

In [5]: data_responses.head(3)

Out[5]:

	Unnamed: 0	Survey Date	Name	User Id	Email	Rating	Classification	Comment
0	0	2020- 08-10	Johanna Vargas T	400842393092	vhannyt@gmail.com	7	passive	NaN
1	1	2020- 08-10	Maria Carolina Parra Rincón	400932763371	mariacarolinaparrar@gmail.com	1	detractor	Atender a las reclamaciones a tiempo para evit
2	2	2020- 08-10	Luz Marina González Pulido	400646948972	14a793667beb4637bc67b25241ee1150@ct.vtex.com.br	0	detractor	No leen con atención, por favor dictar capacit
4								>

Understanding repeated tickets Id

```
In [6]:
       print('there are ' + str(len(data satisfaction['Ticket Id'].unique())) + ' unique tickets')
        print('there are ' + str(len(data satisfaction)) + ' tickets')
        duplicate_tickets = data_satisfaction.groupby('Ticket Id').size().sort_values(ascending=False).reset_index(na
        me ='tickets count')
        duplicate example = data satisfaction[data satisfaction['Ticket Id'] == duplicate tickets['Ticket Id'][3]]
        print('----')
        print('----')
        print('Example of duplicate Tickets')
        duplicate example
        there are 2657 unique tickets
        there are 2774 tickets
        Example of duplicate Tickets
Out[6]:
```

	Unnamed: 0	Requester	User Id	Email	Ticket Id	Brand	Group	Assignee	Satisfaction	Con
2254	2254	leidy huertas	400578886512	leidy.kari96@gmail.com	583147	OFFCORSS	Soporte OFFCORSS	CTS Transporte	good	BUENC Ql RESPONDI UN TIEMI
2255	2255	leidy huertas	400578886512	leidy.kari96@gmail.com	583147	OFFCORSS	Soporte OFFCORSS	CTS Transporte	good	BUENC Ql RESPONDI UN TIEMI
2256	2256	leidy huertas	400578886512	leidy.kari96@gmail.com	583147	OFFCORSS	Soporte OFFCORSS	CTS Transporte	good	
4										>

The repeated ticket Id could be for repeated comments, or for tracking the steps in a request. Therefore are eliminated the repeated tickets Ids with repeated comments.

```
In [7]: data satisfaction.drop duplicates(subset=['Ticket Id', 'processed comment'], inplace = True)
```

Missing values analysis. It is found the percentage of missing comments, and they are replaced with empty values.

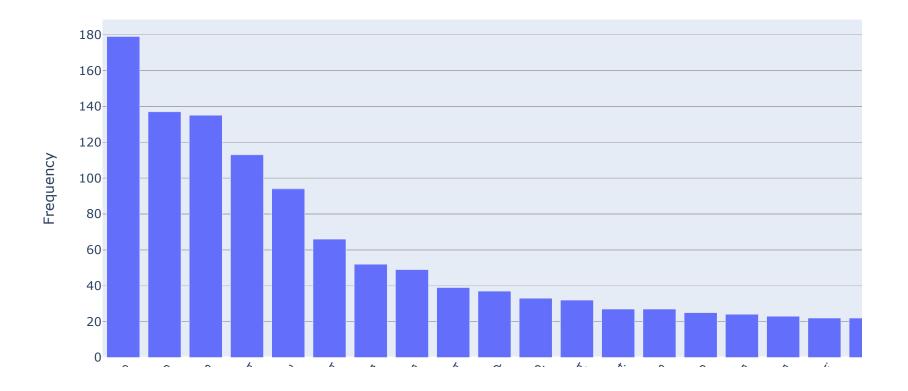
Analysis of comments by groups

```
In [10]: # Select comments by classification
    satisfaction_good = data_satisfaction[(data_satisfaction['Satisfaction'] == 'good')]['processed comment']
    satisfaction_bad = data_satisfaction[(data_satisfaction['Satisfaction'] == 'bad')]['processed comment']
    responses_promoter = data_responses[(data_responses['Classification'] == 'promoter')]['processed comment']
    responses_passive = data_responses[(data_responses['Classification'] == 'passive')]['processed comment']
    responses_detractor = data_responses[(data_responses['Classification'] == 'detractor')]['processed comment']
```

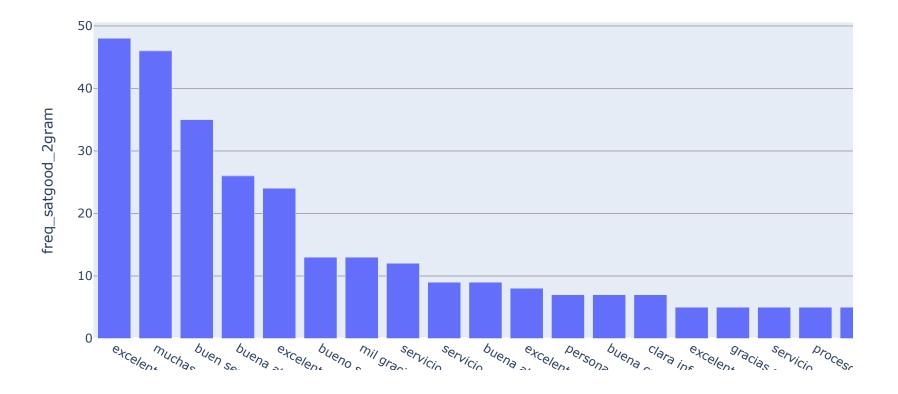
Data Satisfaction good-bad N-grams analysis

In [11]: # Use the get top n words function from data exploring functions (Natesh function) to get the most frequent n -arams # 1- gram for satisfaction good common words = get top n words(satisfaction good, 20,1) grams df = pd.DataFrame(common words, columns = ['word satgood 1gram', 'freq satgood 1gram']) fig = px.bar(grams df, x = 'word satgood 1gram', y = 'freq satgood 1gram', title='Top 20 1-gram from good sat isfaction comments', labels={'freq satgood 1gram':'Frequency', 'word satgood 1gram':'1-gram'}) fig.show() # 2- gram for satisfaction good common words = get top n words(satisfaction good, 20,2) grams_df['word_satgood_2gram'] = [tuple_word[0] for tuple word in common words] grams df['freq satgood 2gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word satgood 2gram', y = 'freq satgood 2gram', title='Top 20 2-grams from good sa tisfaction comments', labels={'word satgood 2gram':'Frequency', 'word satgood 2gram':'2-gram'}) fig.show() # 3- gram for satisfaction good common words = get top n words(satisfaction good, 20,3) grams_df['word_satgood_3gram'] = [tuple_word[0] for tuple_word in common_words] grams df['freq satgood 3gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word satgood 3gram', y = 'freq satgood 3gram', title='Top 20 3-grams from good sa tisfaction comments', labels={'word_satgood_3gram':'Frequency', 'word satgood 3gram':'3-gram'}) fig.show()

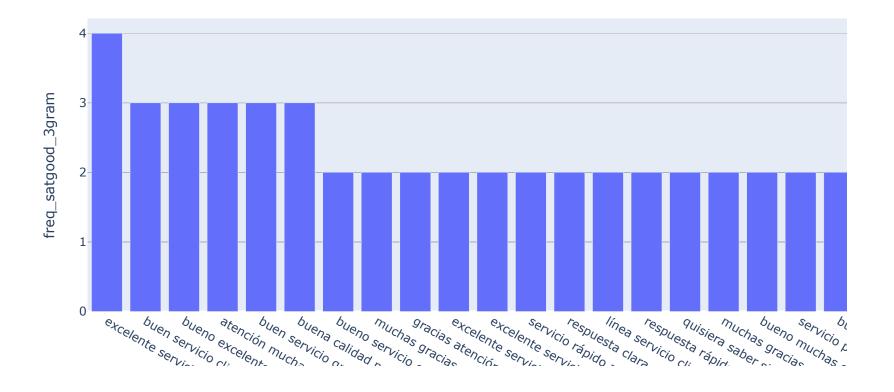
Top 20 1-gram from good satisfaction comments



Top 20 2-grams from good satisfaction comments

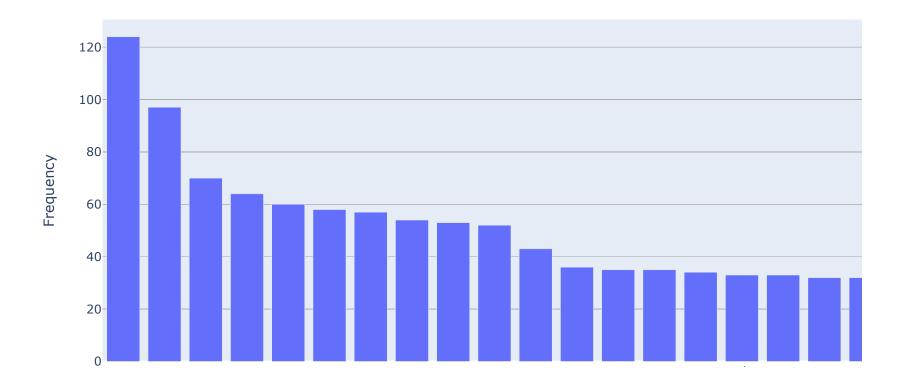


Top 20 3-grams from good satisfaction comments

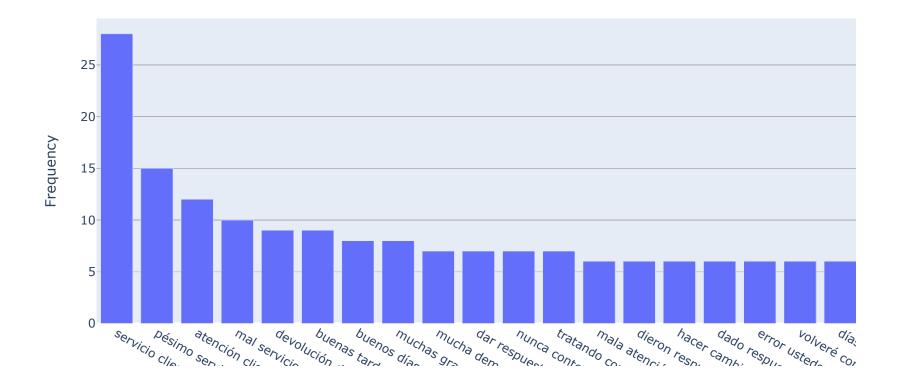


In [12]: # Use the get top n words function from data exploring functions (Natesh function) to get the most frequent n -arams # 1- gram for satisfaction bad common_words = get_top_n_words(satisfaction bad, 20,1) grams df['word satbad 1gram'] = [tuple word[0] for tuple word in common words] grams df['freq satbad 1gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word satbad 1gram', y = 'freq satbad 1gram', title='Top 20 1-gram from bad satisf action comments', labels={'freq satbad 1gram':'Frequency', 'word satbad 1gram':'1-gram'}) fig.show() # 2- gram for satisfaction bad common words = get top n words(satisfaction bad, 20,2) grams df['word satbad 2gram'] = [tuple word[0] for tuple word in common words] grams df['freq satbad 2gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams_df, x = 'word_satbad_2gram', y = 'freq_satbad_2gram', title='Top 20 2-grams from bad satis faction comments', labels={'freq satbad 2gram':'Frequency', 'word satbad 2gram':'2-gram'}) fig.show() # 3- gram for satisfaction bad common words = get top n words(satisfaction bad, 20,3) grams df['word satbad 3gram'] = [tuple word[0] for tuple word in common words] grams df['freq satbad 3gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word satbad 3gram', y = 'freq satbad 3gram', title='Top 20 3-grams from bad satis faction comments', labels={'freq satbad 3gram':'Frequency', 'word satbad 3gram':'3-gram'}) fig.show()

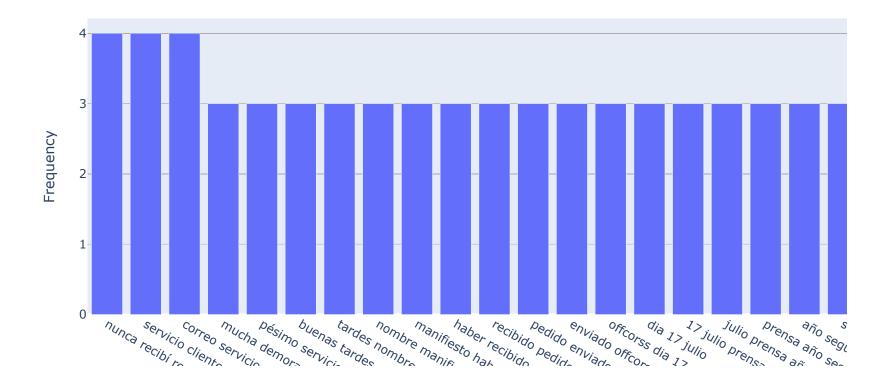
Top 20 1-gram from bad satisfaction comments



Top 20 2-grams from bad satisfaction comments



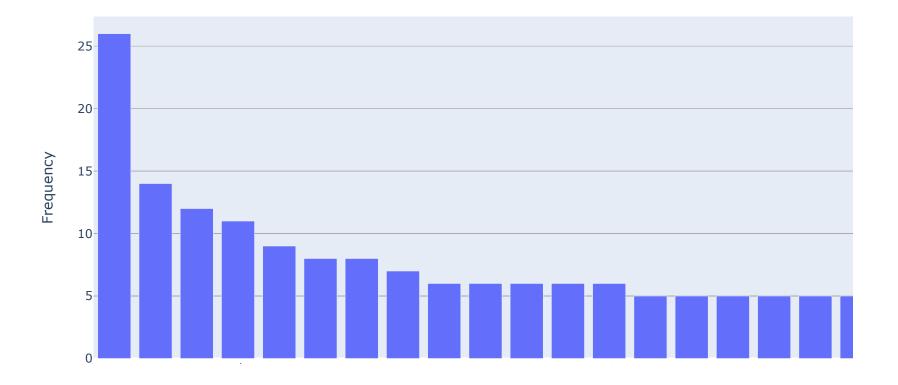
Top 20 3-grams from bad satisfaction comments



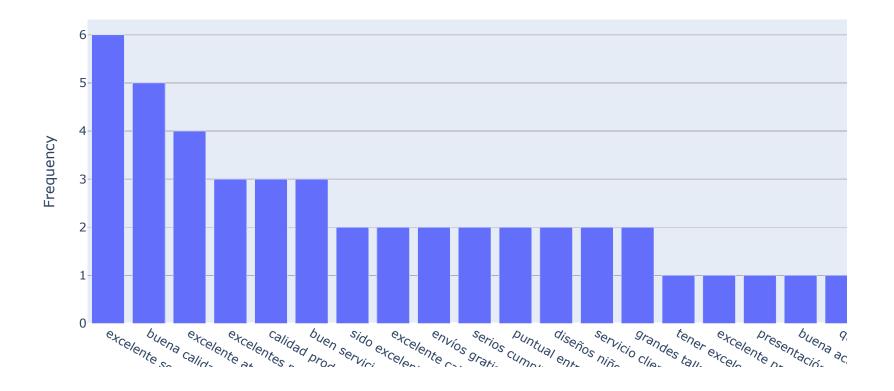
Data Response promoter-passive-detractor N-grams analysis

In [13]: # Use the get top n words function from data exploring functions (Natesh function) to get the most frequent n -arams # 1- gram for response promoter common words = get top n words(responses promoter, 20,1) grams df['word respromoter 1gram'] = [tuple word[0] for tuple word in common words] grams df['freq respromoter 1gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word respromoter 1gram', y = 'freq respromoter 1gram', title='Top 20 1-gram from promoter in response data comments', labels={'freq respromoter 1gram':'Frequency', 'word respromoter 1gram':'1-gram'}) fig.show() # 2- gram for response promoter common words = get top n words(responses promoter, 20,2) grams_df['word_respromoter_2gram'] = [tuple_word[0] for tuple word in common words] grams_df['freq_respromoter_2gram'] = [tuple_word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word respromoter 2gram', y = 'freq respromoter 2gram', title='Top 20 2-grams from promoter response data comments', labels={'freq_respromoter_2gram':'Frequency', 'word_respromoter_2gram':'2-gram'}) fig.show() # 3- gram for response promoter common words = get top n words(responses promoter, 20,3) grams df['word respromoter 3gram'] = [tuple word[0] for tuple word in common words] grams df['freq respromoter 3gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word respromoter 3gram', y = 'freq respromoter 3gram', title='Top 20 3-grams from promoter response data comments', labels={'freq respromoter 3gram':'Frequency', 'word respromoter 3gram':'3-gram'}) fig.show()

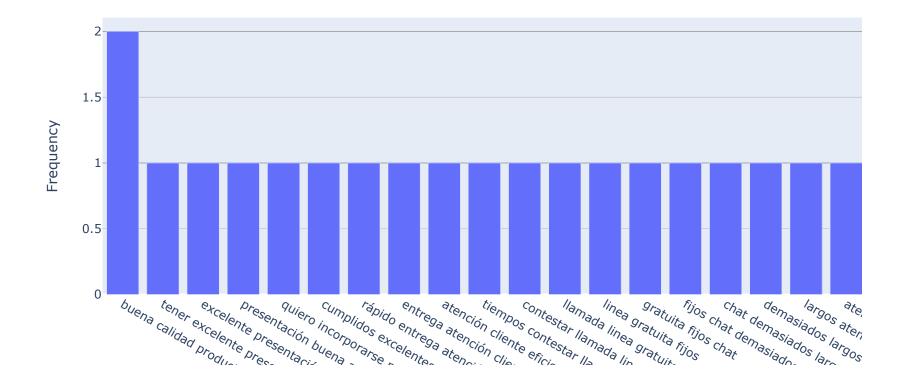
Top 20 1-gram from promoter in response data comments



Top 20 2-grams from promoter response data comments

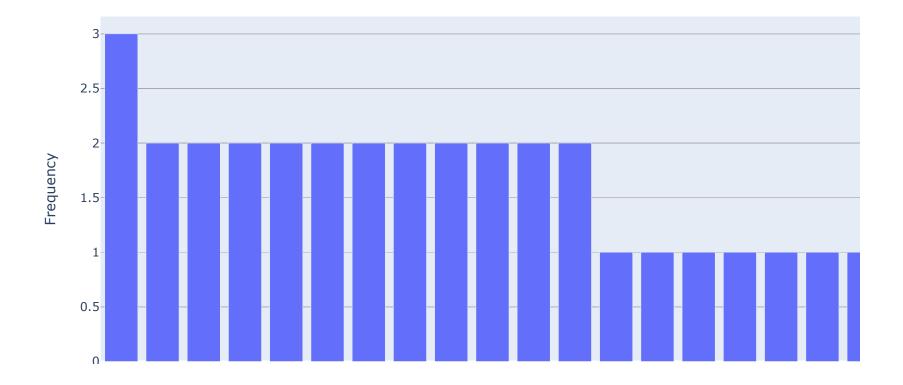


Top 20 3-grams from promoter response data comments

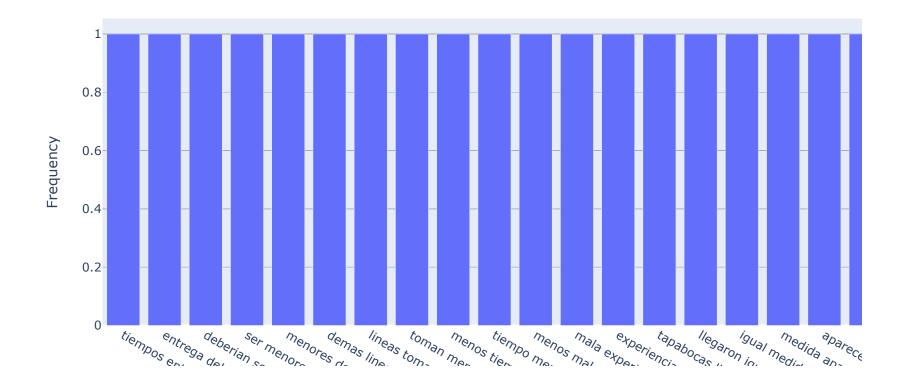


In [14]: # Use the get top n words function from data_exploring_functions (Natesh function) to get the most frequent n -arams # 1- gram for response passive common words = get top n words(responses passive, 20,1) grams df['word respassive 1gram'] = [tuple word[0] for tuple word in common words] grams df['freq respassive 1gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word respassive 1gram', y = 'freq respassive 1gram', title='Top 20 1-gram from pa ssive in response data comments', labels={'freq respassive 1gram':'Frequency', 'word respassive 1gram':'1-gram'}) fig.show() # 2- gram for response passive common words = get top n words(responses passive, 20,2) grams_df['word_respassive_2gram'] = [tuple_word[0] for tuple word in common words] grams df['freq respassive 2gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word respassive 2gram', y = 'freq respassive 2gram', title='Top 20 2-grams from p assive response data comments', labels={'freq respassive 2gram':'Frequency', 'word respassive 2gram':'2-gram'}) fig.show() # 3- gram for response passive common words = get top n words(responses passive, 20,3) grams df['word respassive 3gram'] = [tuple word[0] for tuple word in common words] grams df['freq respassive 3gram'] = [tuple word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word respassive 3gram', y = 'freq respassive 3gram', title='Top 20 3-grams from p assive response data comments', labels={'freq respassive 3gram':'Frequency', 'word respassive 3gram':'3-gram'}) fig.show()

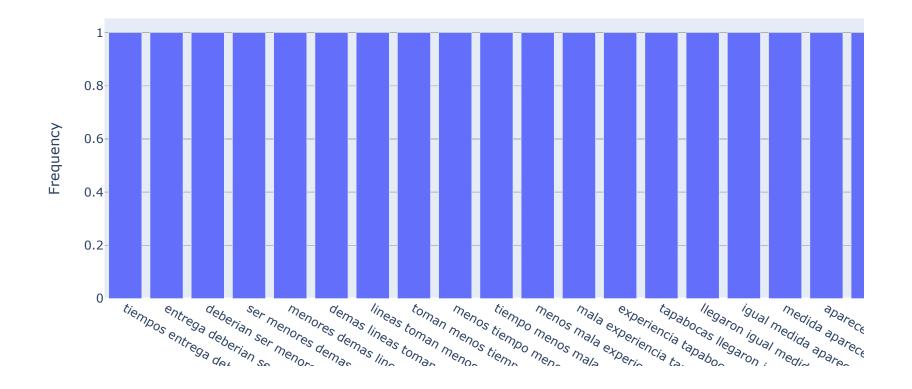
Top 20 1-gram from passive in response data comments



Top 20 2-grams from passive response data comments

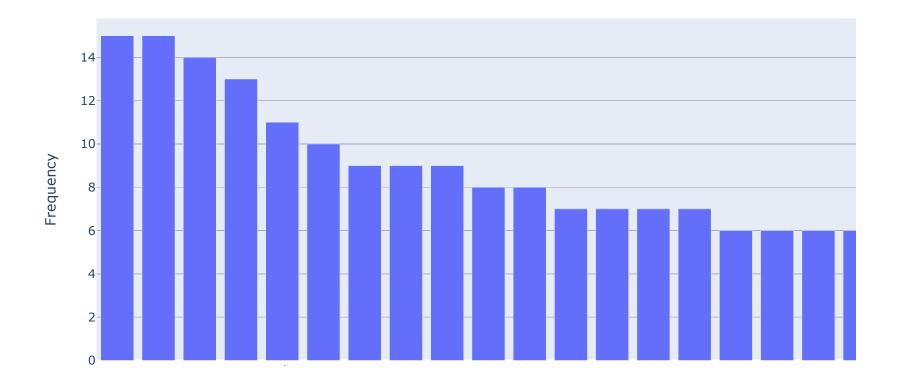


Top 20 3-grams from passive response data comments

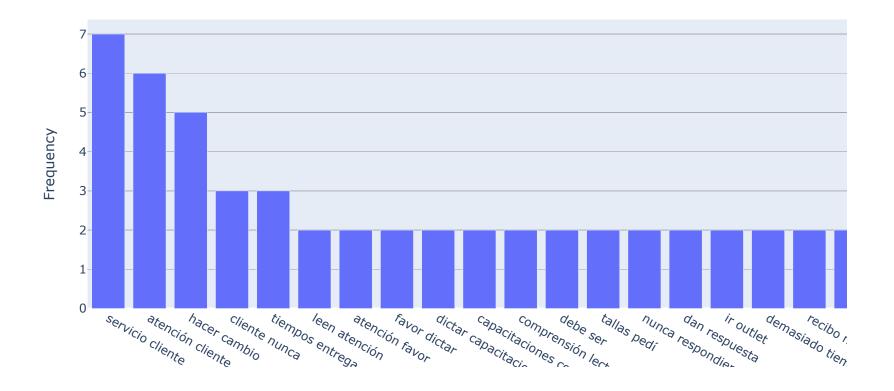


In [15]: |# Use the get top n words function from data exploring functions (Natesh function) to get the most frequent n -arams # 1- gram for response detractor common_words = get_top_n_words(responses detractor, 20,1) grams df['word resdetractor 1gram'] = [tuple word[0] for tuple word in common words] grams_df['freq_resdetractor_1gram'] = [tuple_word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word resdetractor 1gram', y = 'freq resdetractor 1gram', title='Top 20 1-gram fro m detractor in response data comments', labels={'freq resdetractor 1gram':'Frequency', 'word resdetractor 1gram':'1-gram'}) fig.show() # 2- gram for response detractor common words = get top n words(responses detractor, 20,2) grams df['word resdetractor 2gram'] = [tuple word[0] for tuple word in common words] grams_df['freq_resdetractor_2gram'] = [tuple_word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word resdetractor 2gram', y = 'freq resdetractor 2gram', title='Top 20 2-grams fr om detractor response data comments', labels={'freq resdetractor 2gram':'Frequency', 'word resdetractor 2gram':'2-gram'}) fig.show() # 3- gram for response detractor common words = get top n words(responses detractor, 20,3) grams df['word resdetractor 3gram'] = [tuple word[0] for tuple word in common words] grams_df['freq_resdetractor_3gram'] = [tuple_word[1] for tuple word in common words] fig = px.bar(grams df, x = 'word resdetractor 3gram', y = 'freq resdetractor 3gram', title='Top 20 3-grams fr om detractor response data comments', labels={'freq resdetractor 3gram':'Frequency', 'word resdetractor 3gram':'3-gram'}) fig.show()

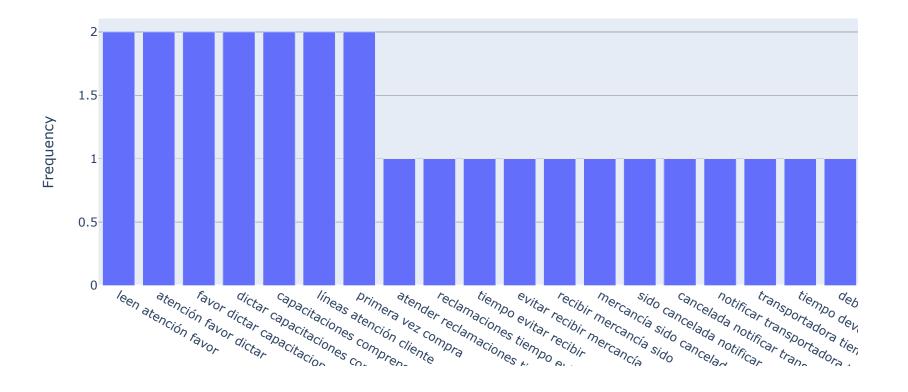
Top 20 1-gram from detractor in response data comments



Top 20 2-grams from detractor response data comments



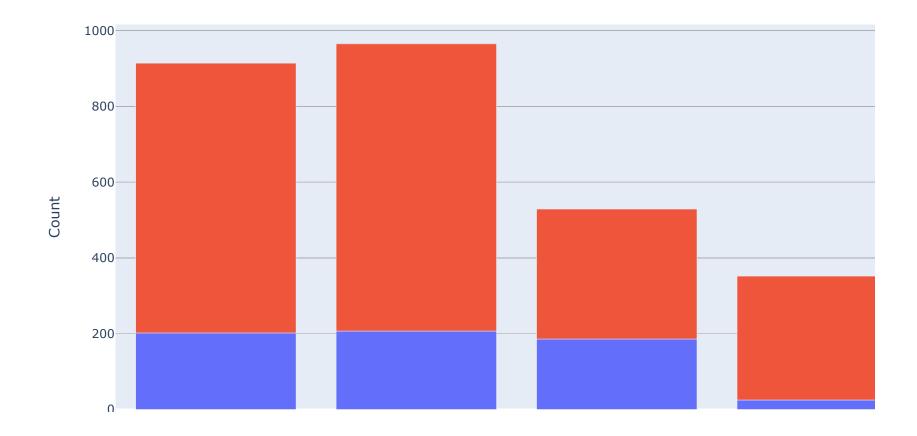
Top 20 3-grams from detractor response data comments



```
In [16]: # Saving the n-grams dataframe.
grams_df.to_csv(os.path.join(base_datapath,'grams.csv'))
```

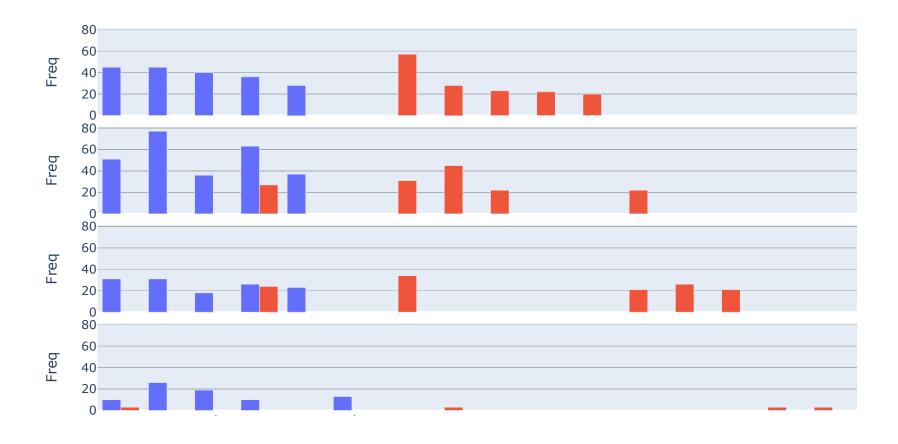
Analysis by organization groups

```
In [17]: group_count = data_satisfaction.groupby(['Group','Satisfaction'])['Ticket Id'].count().reset_index(name='Coun
t')
px.bar(group_count, x ='Group', y= 'Count', color = 'Satisfaction')
```



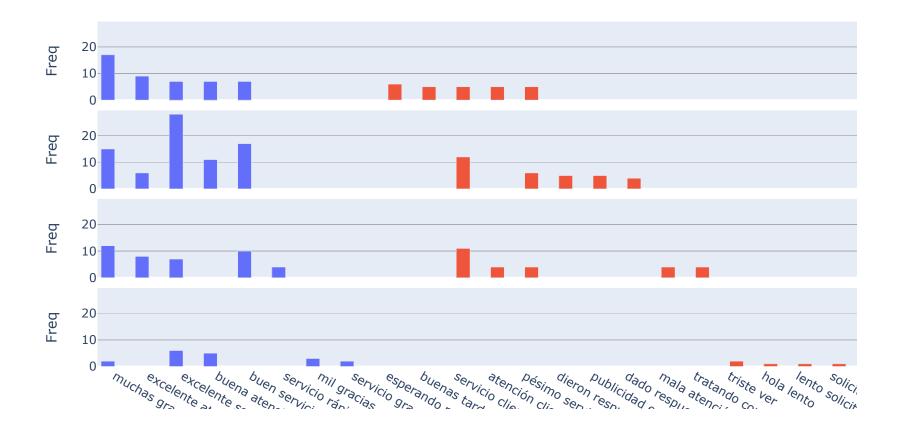
Finding the most frequently 1-grams by organization groups

```
In [18]:
         uniq groups = data satisfaction.Group.unique()
         n = 5
         words group = []
         for group in uniq groups:
             common_words = get_top_n_words(data_satisfaction[(data_satisfaction['Group'] == group) & (data_satisfacti
         on['Satisfaction'] == 'good')]['processed comment'], n, 1)
             for word in common words:
                 words group.append([word[0], word[1], group, 'good'])
             common words = get top n words(data satisfaction[(data satisfaction['Group'] == group) & (data satisfacti
         on['Satisfaction'] == 'bad')]['processed comment'], n, 1)
             for word in common words:
                 words group.append([word[0], word[1], group, 'bad'])
         words 1gram group = pd.DataFrame(words group, columns = ['1gram', 'Freq', 'Group', 'Satisfaction'])
         conver dict = {'Call Center': 'cc', 'Soporte OFFCORSS': 'sop', 'Tienda Virtual' : 'tvir', 'Venta Directa': 'v
         dir'}
         words 1gram group['id'] = words 1gram group['Group'].apply(lambda x: conver dict[x])
         fig = px.bar(words 1gram group, x = '1gram', y = 'Freq', color = 'Satisfaction', facet row='id', barmode='gro
         up')
         fig.show()
         words 1gram group.to csv(os.path.join(base datapath, '1gram organizationgroups.csv'))
```



Finding the most frequently 2-grams by organization groups

```
In [19]:
         uniq groups = data satisfaction.Group.unique()
         n = 5
         words group = []
         for group in uniq groups:
             common_words = get_top_n_words(data_satisfaction[(data_satisfaction['Group'] == group) & (data_satisfacti
         on['Satisfaction'] == 'good')]['processed comment'], n, 2)
             for word in common words:
                 words group.append([word[0], word[1], group, 'good'])
             common words = get top n words(data satisfaction[(data satisfaction['Group'] == group) & (data satisfacti
         on['Satisfaction'] == 'bad')]['processed comment'], n, 2)
             for word in common words:
                 words group.append([word[0], word[1], group, 'bad'])
         words 2gram group = pd.DataFrame(words group, columns = ['2gram', 'Freq', 'Group', 'Satisfaction'])
         conver dict = {'Call Center': 'cc', 'Soporte OFFCORSS': 'sop', 'Tienda Virtual' : 'tvir', 'Venta Directa': 'v
         dir'}
         words 2gram group['id'] = words 1gram group['Group'].apply(lambda x: conver dict[x])
         fig = px.bar(words 2gram group, x = '2gram', y = 'Freq', color = 'Satisfaction', facet row='id', barmode='gro
         up')
         fig.show()
         words 2gram group.to csv(os.path.join(base datapath,'2gram organizationgroups.csv'))
```



Time analysis

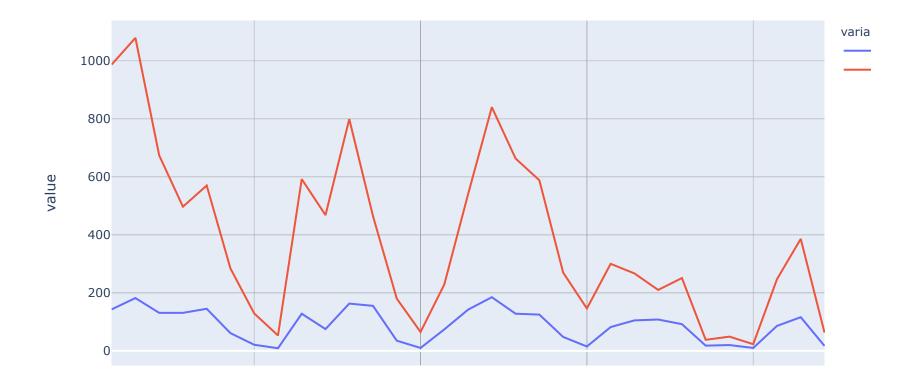
Now we plot the number of comments and words per day in the data satisfaction data.

In [20]: data_satisfaction['datetime'] = pd.to_datetime(data_satisfaction['Survey Date'])
 data_satisfaction['words'] = data_satisfaction['processed comment'].apply(lambda x : len(nltk.word_tokenize(x)))
 comments_day = data_satisfaction.groupby(by = data_satisfaction['datetime'].dt.to_period("D"))[['Ticket Id', 'words']].agg({'Ticket Id':'count', 'words':'sum'}).reset_index().rename(columns={'Ticket Id': 'Count of Comments', 'words': 'Count of words'})
 fig = px.line(comments_day, y = ['Count of Comments', 'Count of words'], x = comments_day['datetime'].dt.to_t imestamp(), title = 'Comments and words over time')
 fig.show()

C:\Users\henry\Anaconda3\envs\ecogeo_tool\lib\site-packages\pandas\core\arrays\datetimes.py:1102: UserWarnin
g:

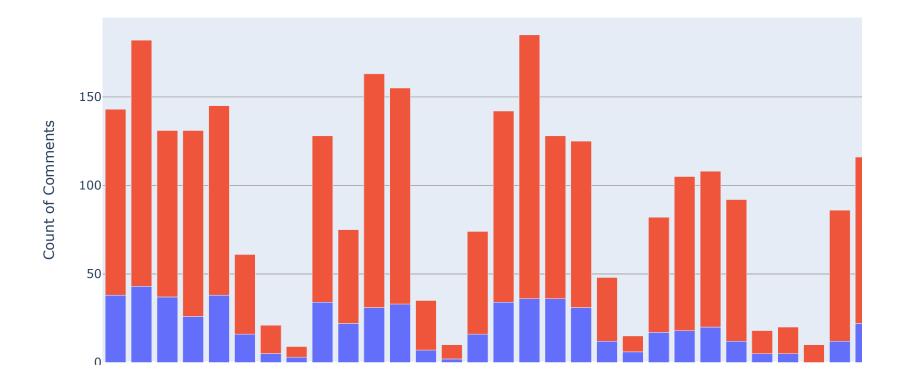
Converting to PeriodArray/Index representation will drop timezone information.

Comments and words over time



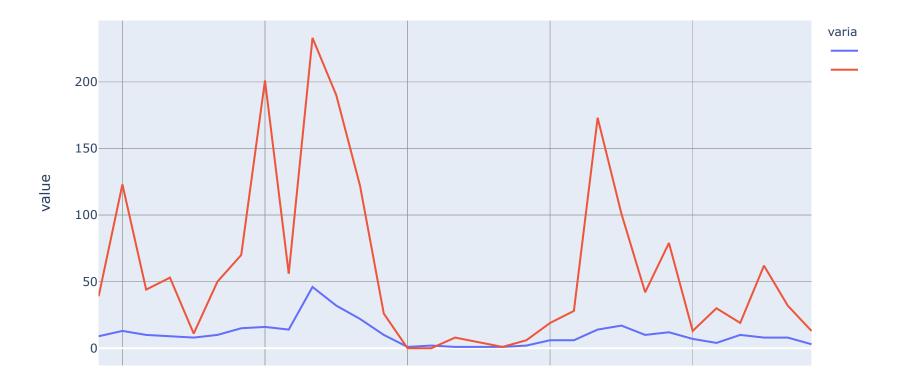
And we plot the good and bad comments per day

Comments over time



In [22]: data_responses['datetime'] = pd.to_datetime(data_responses['Response Date'])
 data_responses['words'] = data_responses['processed comment'].apply(lambda x : len(nltk.word_tokenize(x)))
 comments_day = data_responses.groupby(by = data_responses['datetime'].dt.to_period("H"))[['Unnamed: 0', 'Ratin
 g', 'words']].agg({'Unnamed: 0':'count', 'Rating' : 'mean', 'words':'sum'}).reset_index().rename(columns={'Unna
 med: 0': 'Count of Comments', 'words': 'Count of words'})
 fig = px.line(comments_day, y = ['Count of Comments', 'Count of words'], x = comments_day['datetime'].dt.to_ti
 mestamp(), title = 'Behavior over time')
 fig.show()
 fig = px.line(comments_day, y = ['Rating'], x = comments_day['datetime'].dt.to_timestamp(), title = 'Rating o
 ver time')
 fig.show()
 fig = px.histogram(comments_day, x = ['Rating'], nbins = 50, title = 'Histogram of rating')
 fig.show()

Behavior over time



Rating over time



Histogram of rating

