Define Problem

Adease is an ads and marketing company helping businesses elicit maximum clicks.

Our objective is to forecast or predict the views of different languages for the wikipedia articles

Import dataset, check structure & characteristics

```
import pandas as pd
import numpy as np

from google.colab import drive
drive.mount('/content/drive')

    Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

train_1=pd.read_csv('/content/drive/MyDrive/Colab Notebooks/DS & ML/Projects/13. Adease/train_1.csv')
train_1
```

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	•••	2016- 12-22	2016- 12-23
0	2NE1_zh.wikipedia.org_all-access_spider	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0		32.0	63.0
1	2PM_zh.wikipedia.org_all-access_spider	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0		17.0	42.0
2	3C_zh.wikipedia.org_all-access_spider	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0		3.0	1.0
3	4minute_zh.wikipedia.org_all-access_spider	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0		32.0	10.0
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	NaN		48.0	9.0								
145058	Underworld_(serie_de_películas)_es.wikipedia.o	NaN		NaN	NaN								
145059	Resident_Evil:_Capítulo_Final_es.wikipedia.org	NaN		NaN	NaN								
145060	Enamorándome_de_Ramón_es.wikipedia.org_all-acc	NaN		NaN	NaN								
145061	Hasta_el_último_hombre_es.wikipedia.org_all-ac	NaN		NaN	NaN								
145062	$Francisco_el_matem\'atico_(serie_de_televisi\'on_d$	NaN		NaN	NaN								

145063 rows × 551 columns



train_1.shape (145063, 551)

The shape of train data is around 145063 rows and 551 columns

The data set contains data of about 145063 pages and the respective views of that page from date July 1 2015 to Dec 31 2016.

The values of the views are in objective type convert to float type

```
train_1.iloc[:,1:]=train_1.iloc[:,1:].astype('float64')
```

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	•••	2016- 12-22	2016- 12-23
0	2NE1_zh.wikipedia.org_all-access_spider	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0		32.0	63.0
1	2PM_zh.wikipedia.org_all-access_spider	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0		17.0	42.0
2	3C_zh.wikipedia.org_all-access_spider	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0		3.0	1.0
3	4minute_zh.wikipedia.org_all-access_spider	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0		32.0	10.0
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	NaN		48.0	9.0								
145058	Underworld_(serie_de_películas)_es.wikipedia.o	NaN		NaN	NaN								
145059	Resident_Evil:_Capítulo_Final_es.wikipedia.org	NaN		NaN	NaN								
145060	Enamorándome_de_Ramón_es.wikipedia.org_all-acc	NaN		NaN	NaN								
145061	Hasta_el_último_hombre_es.wikipedia.org_all-ac	NaN		NaN	NaN								
145062	Francisco_el_matemático_(serie_de_televisión_d	NaN		NaN	NaN								
145063 rd	ows × 551 columns												
*													

()+

 $\label{local_policy} $$\exp_1=pd.read_csv('/content/drive/MyDrive/Colab Notebooks/DS \& ML/Projects/13. Adease/Exog_Campaign_eng') $$exog_1$$$



550 rows × 1 columns

```
train_1.isna().sum(axis=1)
```

```
0
            0
0
1
2
            0
3
            0
4
          291
145058
          544
145059
          550
145060
          550
145061
          550
145062
          550
Length: 145063, dtype: int64
```

no. of null values per each page and max is 550 for last 4 pages of dataset

train_1.isna().sum(axis=0),train_1.isna().sum(axis=0).argmax()

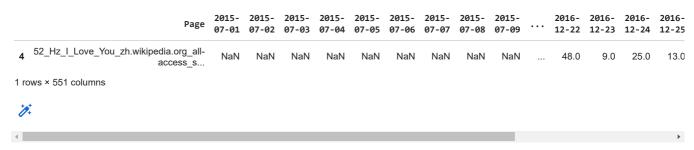
```
(Page 0
2015-07-01 20740
2015-07-02 20816
2015-07-03 20544
2015-07-04 20654
...
2016-12-27 3701
2016-12-28 3822
```

```
2016-12-29 3826
2016-12-30 3635
2016-12-31 3465
Length: 551, dtype: int64, 2)
```

no of null values per each date and max is having with date 2015-07-02

Impute the null values using linear interpolation

train_1.iloc[[4]]



Null values indicate either there might not be any views or page might not be in existence or created by then

So impute the values either with 0 or iterpolate

if we eliminate rows even with a single value

train_1[~(train_1.isna().any(axis=1))]

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	 2016- 12-22	201(12-:
0	2NE1_zh.wikipedia.org_all-access_spider	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0	 32.0	63
1	2PM_zh.wikipedia.org_all-access_spider	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0	 17.0	42
2	3C_zh.wikipedia.org_all-access_spider	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0	 3.0	1
3	4minute_zh.wikipedia.org_all-access_spider	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0	 32.0	10
5	5566_zh.wikipedia.org_all-access_spider	12.0	7.0	4.0	5.0	20.0	8.0	5.0	17.0	24.0	 16.0	27
144944	Chichén_Itzá_es.wikipedia.org_all-access_spider	8.0	13.0	19.0	14.0	6.0	5.0	10.0	9.0	5.0	 15.0	18
144945	Fecundación_es.wikipedia.org_all-access_spider	29.0	16.0	6.0	11.0	33.0	4.0	11.0	16.0	10.0	 8.0	8
144946	Gran_Hermano_VIP_(España)_es.wikipedia.org_all	4.0	25.0	7.0	11.0	6.0	6.0	16.0	11.0	23.0	 12.0	299
144947	Modelo_atómico_de_Thomson_es.wikipedia.org_all	0.0	2.0	6.0	6.0	7.0	5.0	4.0	6.0	7.0	 13.0	1
144948	Copa_América_2019_es.wikipedia.org_all-access	3.0	10.0	41.0	17.0	16.0	14.0	8.0	12.0	4.0	 8.0	8
117277 rc	owe x 551 columns											

117277 rows × 551 columns



117277/145063

0.808455636516548

117277+27786

145063

20% of data loss

import math

train_1.iloc[1,2]

14.0

null_rows=train_1[(train_1.isna().any(axis=1))]
null_rows

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	 2016- 12-22	2016- 12-23
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	NaN	 48.0	9.0								
6	91Days_zh.wikipedia.org_all-access_spider	NaN	 2.0	7.0								
10	ASTRO_zh.wikipedia.org_all-access_spider	NaN	NaN	NaN	NaN	NaN	1.0	1.0	NaN	NaN	 11.0	38.0
13	AlphaGo_zh.wikipedia.org_all-access_spider	NaN	 14.0	13.0								
19	B-PROJECT_zh.wikipedia.org_all-access_spider	NaN	 4.0	26.0								
145058	Underworld_(serie_de_películas)_es.wikipedia.o	NaN	 NaN	NaN								
145059	Resident_Evil:_Capítulo_Final_es.wikipedia.org	NaN	 NaN	NaN								
145060	Enamorándome_de_Ramón_es.wikipedia.org_all-acc	NaN	 NaN	NaN								
145061	Hasta_el_último_hombre_es.wikipedia.org_all-ac	NaN	 NaN	NaN								
145062	$Francisco_el_matem\'atico_(serie_de_televisi\'on_d$	NaN	 NaN	NaN								
27786 rov	vs × 551 columns											
%												

Impute the values with interpolation

train_1.iloc[:,1:]

4

	2015- 07-01	2015- 07-02	2015- 07-03		2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09		•••	2016- 12-22	2016- 12-23	2016- 12-24		2016- 12-26	2016- 12-27	2016- 12-28	_
0	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0	24.0		32.0	63.0	15.0	26.0	14.0	20.0	22.0	
1	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0	4.0		17.0	42.0	28.0	15.0	9.0	30.0	52.0	
2	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0	4.0		3.0	1.0	1.0	7.0	4.0	4.0	6.0	
3	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0	16.0		32.0	10.0	26.0	27.0	16.0	11.0	17.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		48.0	9.0	25.0	13.0	3.0	11.0	27.0	
145058	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN	NaN	13.0	12.0	13.0	
145059	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN	NaN	NaN	
145060	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN	NaN	NaN	
145061	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN	NaN	NaN	
145062	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN	NaN	NaN	

145063 rows × 550 columns



##Linear interpolation

train_1_after_interpolation=train_1.iloc[:,1:].interpolate(method='linear',axis=1,limit_direction='both')

 ${\tt train_1_after_interpolation}$

					2015- 07-05					2015- 07-10	•••						2016- 12-27		_
0	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0	24.0		32.0	63.0	15.0	26.0	14.0	20.0	22.0	
1	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0	4.0		17.0	42.0	28.0	15.0	9.0	30.0	52.0	
2	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0	4.0		3.0	1.0	1.0	7.0	4.0	4.0	6.0	
3	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0	16.0		32.0	10.0	26.0	27.0	16.0	11.0	17.0	
4	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0		48.0	9.0	25.0	13.0	3.0	11.0	27.0	
145058	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0		13.0	13.0	13.0	13.0	13.0	12.0	13.0	
145050	NaN	NaN	NaN	NaNi	NaNi	NaN	NaN	NaNi	NaN	NaNi		NaN	NaNi	NaN	NaN	NaN	NaNi	NaN	

 $train_1_after_interpolation=train_1[['Page']].join(train_1_after_interpolation)$

 ${\tt train_1_after_interpolation}$

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	•••	2016- 12-22	2016- 12-23
0	2NE1_zh.wikipedia.org_all-access_spider	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0		32.0	63.0
1	2PM_zh.wikipedia.org_all-access_spider	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0		17.0	42.0
2	3C_zh.wikipedia.org_all-access_spider	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0		3.0	1.0
3	4minute_zh.wikipedia.org_all-access_spider	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0		32.0	10.0
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0		48.0	9.0
145058	Underworld_(serie_de_películas)_es.wikipedia.o	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0		13.0	13.0
145059	Resident_Evil:_Capítulo_Final_es.wikipedia.org	NaN		NaN	NaN								
145060	Enamorándome_de_Ramón_es.wikipedia.org_all-acc	NaN		NaN	NaN								
145061	Hasta_el_último_hombre_es.wikipedia.org_all-ac	NaN		NaN	NaN								
145062	Francisco_el_matemático_(serie_de_televisión_d	NaN		NaN	NaN								

145063 rows × 551 columns



 $train_1_after_interpolation:train_1_after_interpolation.isna().any(axis=1))]$

train_1_after_interpolation

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	 2016- 12-22	2016-1
0	2NE1_zh.wikipedia.org_all-access_spider	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0	 32.0	63.0000
1	2PM_zh.wikipedia.org_all-access_spider	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0	 17.0	42.0000
2	3C_zh.wikipedia.org_all-access_spider	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0	 3.0	1.0000
3	4minute_zh.wikipedia.org_all-access_spider	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0	 32.0	10.0000
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	 48.0	9.0000
145054	Skam_(serie_de_televisión)_es.wikipedia.org_al	26.0	26.0	26.0	26.0	26.0	26.0	26.0	26.0	26.0	 8.0	9.0000
145055	Legión_(serie_de_televisión)_es.wikipedia.org	50.0	50.0	50.0	50.0	50.0	50.0	50.0	50.0	50.0	 1.0	2.0000
145056	Doble_tentación_es.wikipedia.org_all-access_sp	11.0	11.0	11.0	11.0	11.0	11.0	11.0	11.0	11.0	 21.0	24.3333
145057	Mi_adorable_maldición_es.wikipedia.org_all-acc	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	 0.0	0.0000
145058	Underworld_(serie_de_películas)_es.wikipedia.o	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0	 13.0	13.0000
144411 rc	ows × 551 columns											

1

```
train\_1\_after\_interpolation=train\_1\_after\_interpolation.isna().any(axis=1))]
```

 ${\tt train_1_after_interpolation}$

	Page	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	•••	2016- 12-22	2016-1
0	2NE1_zh.wikipedia.org_all-access_spider	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0		32.0	63.0000
1	2PM_zh.wikipedia.org_all-access_spider	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0		17.0	42.0000
2	3C_zh.wikipedia.org_all-access_spider	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0		3.0	1.0000
3	4minute_zh.wikipedia.org_all-access_spider	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0		32.0	10.0000
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0		48.0	9.0000
145054	Skam_(serie_de_televisión)_es.wikipedia.org_al	26.0	26.0	26.0	26.0	26.0	26.0	26.0	26.0	26.0		8.0	9.0000
145055	Legión_(serie_de_televisión)_es.wikipedia.org	50.0	50.0	50.0	50.0	50.0	50.0	50.0	50.0	50.0		1.0	2.0000
145056	Doble_tentación_es.wikipedia.org_all-access_sp	11.0	11.0	11.0	11.0	11.0	11.0	11.0	11.0	11.0		21.0	24.3333
145057	Mi_adorable_maldición_es.wikipedia.org_all-acc	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0		0.0	0.0000
145058	Underworld_(serie_de_películas)_es.wikipedia.o	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0	13.0		13.0	13.0000
144411 rc	ows × 551 columns												
7													

144411/145063

0.9955054011015904

As the data constitutes about 99.5% after dropping its ok to move forward

▼ Data Visualization

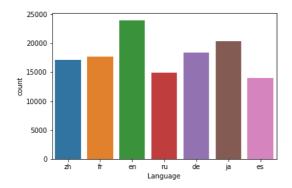
```
trial 1=train 1 after interpolation[train 1 after interpolation['Page'].str.contains('wikipedia.org')]
def split_1(x):
  array_1=[]
  # y=a.split('wikipedia.org')
  # x=list(map(lambda z: z.strip('.'), a))
  array_1.append(x[0][:-3])
  array_1.append(x[0][-3:].strip('.'))
  list_1=x[1].split('_')
array_1.append("_".join(list_1[:-1]))
  array_1.append(list_1[-1])
  return "$".join(array_1)
trial_1.loc[:,'split_1']=trial_1.loc[:,'Page'].apply(lambda x:x.split('wikipedia.org'))
      /usr/local/lib/python3.8/dist-packages/pandas/core/indexing.py:1667: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead
      See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus</a>
        self.obj[key] = value
\label{linear_split} trial_1['array'] = trial_1['split_1'].apply(lambda \ x: \ split_1(x))
      <ipython-input-31-bbc2803ba9f3>:1: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame.
      Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus trial_1['array']=trial_1['split_1'].apply(lambda x: split_1(x))

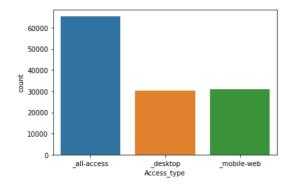
```
4
trial_1['array']
     0
                                           2NE1_$zh$_all-access$spider
     1
                                            {\tt 2PM\_\$zh\$\_all-access\$spider}
     2
                                             3C_$zh$_all-access$spider
      3
                                       4minute_$zh$_all-access$spider
      4
                             52_Hz_I_Love_You_$zh$_all-access$spider
      145054
                 Skam_(serie_de_televisión)_$es$_all-access$spider
      145055
                 Legión_(serie_de_televisión)_$es$_all-access$s...
      145056
                              Doble tentación $es$ all-access$spider
     145057
                       Mi_adorable_maldición_$es$_all-access$spider
                 Underworld_(serie_de_películas)_$es$_all-acces...
     145058
     Name: array, Length: 126683, dtype: object
trial_1[['Title','Language','Access_type','Access_origin']]=trial_1['array'].str.split("$",expand=True)
      /usr/local/lib/python3.8/dist-packages/pandas/core/frame.py:3641: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame.
      Try using .loc[row_indexer,col_indexer] = value instead
      See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus</a>
        self[k1] = value[k2]
trial 1
                                                                                         2015-
                                                                                                         2015-
                                                                                                                 2015-
                                                                                                                         2015-
                                                                                                                                 2015-
                                                                                                                                               2016-
                                                                  2015-
                                                                         2015-
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                                                                                                 2015-
                                                                                                                                                       2016-1
                                                           Page
                                                                  07-01
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                                                                                 07-03
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                                                                                                 07-05
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                                                                                                                 07-07
                                                                                                                         07-08
                                                                                                                                 07-09
                                                                                                                                               12-28
         0
                        2NE1_zh.wikipedia.org_all-access_spider
                                                                   18.0
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                      4minute_zh.wikipedia.org_all-access_spider
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       145058 Underworld_(serie_de_películas)_es.wikipedia.o...
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      126683 rows × 557 columns
       1
trial 1.shape
      (126683, 557)
trial_1['Language'].value_counts()
      en
             24010
             20340
      jа
      de
             18438
             17761
      fr
      zh
             17103
     ru
             14990
             14041
      es
     Name: Language, dtype: int64
```

import seaborn as sns
import matplotlib.pyplot as plt

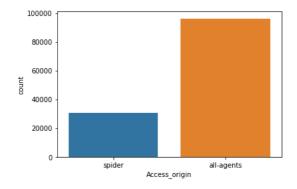
```
sns.countplot(x=trial_1['Language'])
plt.show()
```



sns.countplot(x=trial_1['Access_type'])
plt.show()



```
sns.countplot(x=trial_1['Access_origin'])
plt.show()
```



top 10 days where the visits are more trial_1.columns

trial_1.iloc[:,1:551].sum(axis=0).nlargest(10)

```
2016-08-15
              3.210200e+08
2016-07-26
              3.122488e+08
2016-07-25
              3.118267e+08
2016-08-10
              3.084869e+08
2016-08-14
              3.084316e+08
2016-08-01
              3.083562e+08
2016-07-27
              3.059005e+08
2016-08-12
              3.045786e+08
              3.035776e+08
2016-08-08
2016-11-09
              3.033059e+08
dtype: float64
```

Those were the dates with highest number of visits

trial_2=trial_1.iloc[:,[0,551,552,553,554,555,556]].join(trial_1.iloc[:,1:551]) trial_2

	Page	split_1	array	
0	2NE1_zh.wikipedia.org_all-access_spider	[2NE1_zh., _all-access_spider]	2NE1_\$zh\$_all-access\$spider	
1	2PM_zh.wikipedia.org_all-access_spider	[2PM_zh., _all-access_spider]	2PM_\$zh\$_all-access\$spider	
2	3C_zh.wikipedia.org_all-access_spider	[3C_zh., _all-access_spider]	3C_\$zh\$_all-access\$spider	
3	4minute_zh.wikipedia.org_all-access_spider	[4minute_zh., _all-access_spider]	4minute_\$zh\$_all-access\$spider	
4	52_Hz_I_Love_You_zh.wikipedia.org_all-access_s	[52_Hz_I_Love_You_zh., _all- access_spider]	52_Hz_I_Love_You_\$zh\$_all- access\$spider	
145054	Skam_(serie_de_televisión)_es.wikipedia.org_al	[Skam_(serie_de_televisión)_es., _all-access_s	Skam_(serie_de_televisión)_\$es\$_all- access\$spider	Skam __
145055	Legión_(serie_de_televisión)_es.wikipedia.org	[Legión_(serie_de_televisión)_es., _all-access	Legión_(serie_de_televisión)_\$es\$_all-access\$s	Legión _.
145056	Doble_tentación_es.wikipedia.org_all-access_sp	[Doble_tentación_es., _all- access_spider]	Doble_tentación_\$es\$_all-access\$spider	
145057	Mi_adorable_maldición_es.wikipedia.org_all-acc	[Mi_adorable_maldición_es., _all- access_spider]	Mi_adorable_maldición_\$es\$_all- access\$spider	M
145058	Underworld_(serie_de_películas)_es.wikipedia.o	[Underworld_(serie_de_películas)_es., _all-acc	Underworld_(serie_de_películas)_\$es\$_all-acces	Underworld
126683 rd	ows × 557 columns			
+=+				



 $\hbox{\tt\#\#language wise highest number of visits on which day}$ trial_3=trial_2.iloc[:,3:] trial_3

	Title	Language	Access_type	Access_origin	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	•••	2016- 12-22	26
0	2NE1_	zh	_all-access	spider	18.0	11.0	5.0	13.0	14.0	9.0		32.0	63
1	2PM_	zh	_all-access	spider	11.0	14.0	15.0	18.0	11.0	13.0		17.0	42
2	3C_	zh	_all-access	spider	1.0	0.0	1.0	1.0	0.0	4.0		3.0	1
3	4minute_	zh	_all-access	spider	35.0	13.0	10.0	94.0	4.0	26.0		32.0	10
4	52_Hz_I_Love_You_	zh	_all-access	spider	38.0	38.0	38.0	38.0	38.0	38.0		48.0	9
145054	Skam_(serie_de_televisión)_	es	_all-access	spider	26.0	26.0	26.0	26.0	26.0	26.0		8.0	9
145055	Legión_(serie_de_televisión)_	es	_all-access	spider	50.0	50.0	50.0	50.0	50.0	50.0		1.0	2
145056	Doble_tentación_	es	_all-access	spider	11.0	11.0	11.0	11.0	11.0	11.0		21.0	24
145057	Mi_adorable_maldición_	es	_all-access	spider	21.0	21.0	21.0	21.0	21.0	21.0		0.0	0
145058	Underworld_(serie_de_películas)_	es	_all-access	spider	13.0	13.0	13.0	13.0	13.0	13.0		13.0	13
126683 rd	ows × 554 columns												



([trial_3.columns[1]])+(trial_3.columns[4:]).tolist()

trial_4_columns=([trial_3.columns[1]])+(trial_3.columns[4:]).tolist()

trial_4=trial_3[trial_4_columns] trial_4

	Language	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	2015- 07-07	2015- 07-08	2015- 07-09	•••	2016- 12-22	2016-12- 23	2016-12- 24	2016- 12-25	2016-12- 26	
0	zh	18.0	11.0	5.0	13.0	14.0	9.0	9.0	22.0	26.0		32.0	63.000000	15.000000	26.0	14.000000	2
1	zh	11.0	14.0	15.0	18.0	11.0	13.0	22.0	11.0	10.0		17.0	42.000000	28.000000	15.0	9.000000	3
2	zh	1.0	0.0	1.0	1.0	0.0	4.0	0.0	3.0	4.0		3.0	1.000000	1.000000	7.0	4.000000	
3	zh	35.0	13.0	10.0	94.0	4.0	26.0	14.0	9.0	11.0		32.0	10.000000	26.000000	27.0	16.000000	1
4	zh	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0	38.0		48.0	9.000000	25.000000	13.0	3.000000	1
145054	es	26.0	26.0	26.0	26.0	26.0	26.0	26.0	26.0	26.0		8.0	9.000000	9.000000	19.0	17.000000	
145055	es	50.0	50.0	50.0	50.0	50.0	50.0	50.0	50.0	50.0		1.0	2.000000	1.000000	1.0	3.000000	
145056	es	11.0	11.0	11.0	11.0	11.0	11.0	11.0	11.0	11.0		21.0	24.333333	27.666667	31.0	34.333333	3
145057	es	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0		0.0	0.000000	0.000000	0.0	0.000000	

trial_5=trial_4.groupby(['Language']).agg(lambda x: x.sum()).T.reset_index().set_index('index')

trial_5

Language	de	en	es	fr	ja	ru	zh	1
index								
2015-07-01	1.507832e+07	9.415409e+07	1.618992e+07	9.232359e+06	1.549925e+07	1.170350e+07	5.159275e+06	
2015-07-02	1.489782e+07	9.387991e+07	1.551257e+07	9.286675e+06	1.725639e+07	1.186729e+07	5.165336e+06	
2015-07-03	1.437188e+07	8.960875e+07	1.433925e+07	8.959763e+06	1.594099e+07	1.116309e+07	5.137253e+06	
2015-07-04	1.333821e+07	9.290429e+07	1.351702e+07	9.521535e+06	1.909167e+07	1.063287e+07	5.176900e+06	
2015-07-05	1.521015e+07	9.563976e+07	1.462066e+07	9.362161e+06	1.846234e+07	1.117435e+07	5.454677e+06	
2016-12-27	2.032261e+07	1.458706e+08	1.594582e+07	1.528222e+07	1.626896e+07	1.520168e+07	6.487991e+06	
2016-12-28	1.934974e+07	1.415205e+08	1.657789e+07	1.378210e+07	1.629641e+07	1.416161e+07	6.522969e+06	
2016-12-29	1.864423e+07	1.507996e+08	1.564768e+07	1.340043e+07	1.782839e+07	1.364024e+07	6.051296e+06	
2016-12-30	1.780216e+07	1.256468e+08	1.156067e+07	1.247502e+07	1.959575e+07	1.222803e+07	6.117870e+06	
2016-12-31	1.675861e+07	1.238632e+08	1.107802e+07	1.150493e+07	2.460054e+07	1.338433e+07	6.305259e+06	
550 rows × 7	columns							

trial_5.idxmax(axis=0),trial_5.max(axis=0)

```
(Language
de 2015-12-07
en 2016-07-26
 es 2016-11-09
 fr
        2016-04-24
.. 2016-04-24
ja 2016-01-11
ru 2016-07-20
       2016-07-28
2016-01-16
 zh
dtype: object, Language
de 2.510500e+07
en 2.050938e+08
es 3.045691e+07
fr 2.000300e+07
 ja
        3.289058e+07
       4.523846e+07
        1.175250e+07
 zh
 dtype: float64)
```

The above are the dates for each language on which the views are highest respectively

trial 3

	Title	Language	Access_type	Access_origin	2015- 07-01	2015- 07-02	2015- 07-03	2015- 07-04	2015- 07-05	2015- 07-06	•••	2016- 12-22	2(
0	2NE1_	zh	_all-access	spider	18.0	11.0	5.0	13.0	14.0	9.0		32.0	63
1	2PM_	zh	_all-access	spider	11.0	14.0	15.0	18.0	11.0	13.0		17.0	42
2	3C_	zh	_all-access	spider	1.0	0.0	1.0	1.0	0.0	4.0		3.0	1
3	4minute_	zh	_all-access	spider	35.0	13.0	10.0	94.0	4.0	26.0		32.0	10
4	52_Hz_I_Love_You_	zh	_all-access	spider	38.0	38.0	38.0	38.0	38.0	38.0		48.0	9
145054	Skam_(serie_de_televisión)_	es	_all-access	spider	26.0	26.0	26.0	26.0	26.0	26.0		8.0	9
145055	Legión_(serie_de_televisión)_	es	_all-access	spider	50.0	50.0	50.0	50.0	50.0	50.0		1.0	2
145056	Doble_tentación_	es	_all-access	spider	11.0	11.0	11.0	11.0	11.0	11.0		21.0	24
	onao: wona_(oono_ao_ponoaiao)_		_u uooooo	opidoi	10.0	10.0							

 $trial_3.groupby(['Language','Access_type']).agg(lambda~x:~x.sum()~if~x.dtype=='float64'~else~x.head(1)).drop(columns=['Tital_action for the columns for the$

Language	Access_type	
de	_all-access	4.827087e+09
	_desktop	2.277197e+09
	_mobile-web	2.413740e+09
en	_all-access	3.133542e+10
	_desktop	1.833471e+10
	_mobile-web	1.238175e+10
es	_all-access	4.964843e+09
	_desktop	2.285070e+09
	_mobile-web	2.630231e+09
fr	_all-access	3.345472e+09
	_desktop	1.586711e+09
	_mobile-web	1.648578e+09
ja	_all-access	4.493121e+09
	_desktop	2.724364e+09
	_mobile-web	2.646978e+09
ru	_all-access	4.398165e+09
	_desktop	2.760635e+09
	_mobile-web	1.586912e+09
zh	_all-access	1.778901e+09
	_desktop	1.047478e+09
	_mobile-web	6.672629e+08
-d-4 C1	+	

dtype: float64

The above are the views for every langauge and accesse type

```
trial_6=trial_3.groupby(['Language','Access_type']).agg(lambda x: x.sum() if x.dtype=='float64' else x.head(1)).drop(columns=['Title','A
trial_6.columns=['Language','Access_type','Sum']
trial_6
```

```
Language Access_type Sum

0 de _all-access 4.827087e+09

1 de _desktop 2.277197e+09

2 de _mobile-web 2.413740e+09

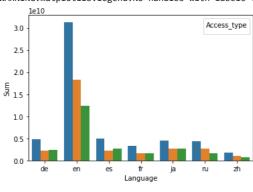
3 en _all-access 3.133542e+10

4 en _desktop 1.833471e+10

5 en mobile-web 1.238175e+10

plt.plot(figsize=(20,10))
sns.barplot(data=trial_6,x='Language',y='Sum',hue='Access_type')
plt.show()
```

 $\label{locality} {\tt WARNING:matplotlib.legend:No\ handles\ with\ labels\ found\ to\ put\ in\ legend.}$



English language is having major visits and as per the access type

checking stationarity

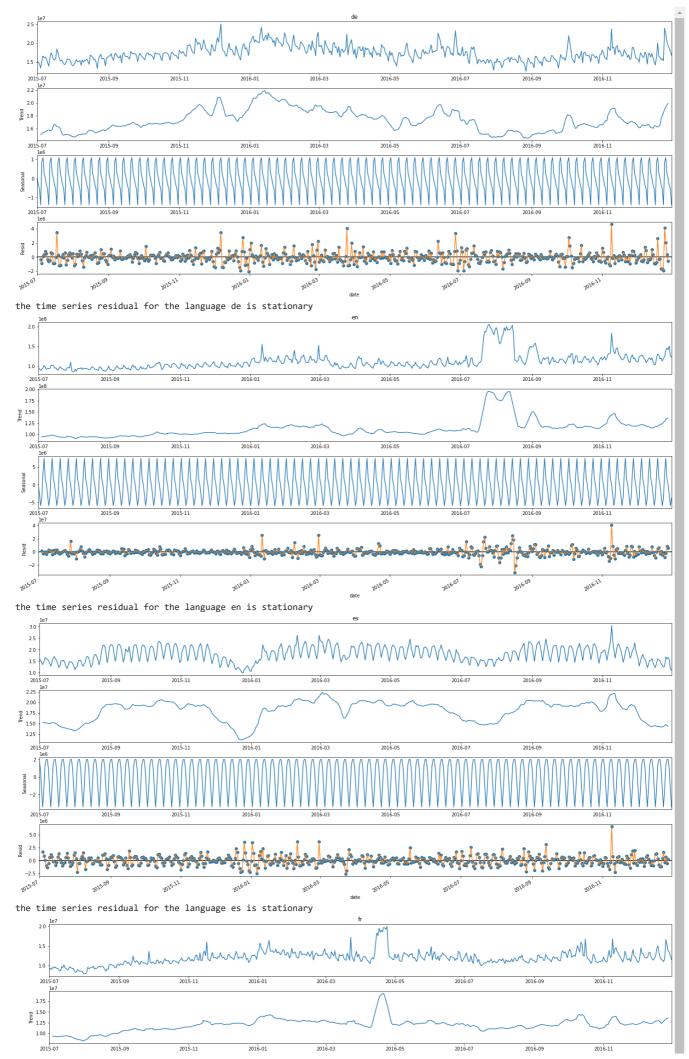
trial_5

Langua	ge	de	en	es	fr	ja	ru	zh	%
inc	ex								
2015-07	01 1.507832e	+07 9.4	15409e+07	1.618992e+07	9.232359e+06	1.549925e+07	1.170350e+07	5.159275e+06	
2015-07	02 1.489782e	+07 9.3	87991e+07	1.551257e+07	9.286675e+06	1.725639e+07	1.186729e+07	5.165336e+06	
2015-07	.03 1.437188e	+07 8.9	60875e+07	1.433925e+07	8.959763e+06	1.594099e+07	1.116309e+07	5.137253e+06	
2015-07	04 1.333821e	+07 9.2	90429e+07	1.351702e+07	9.521535e+06	1.909167e+07	1.063287e+07	5.176900e+06	
2015-07	05 1.521015e	+07 9.5	63976e+07	1.462066e+07	9.362161e+06	1.846234e+07	1.117435e+07	5.454677e+06	
					•••				
2016-12	27 2.032261e	+07 1.4	58706e+08	1.594582e+07	1.528222e+07	1.626896e+07	1.520168e+07	6.487991e+06	
2016-12	28 1.934974e	+07 1.4	15205e+08	1.657789e+07	1.378210e+07	1.629641e+07	1.416161e+07	6.522969e+06	
2016-12	29 1.864423e	+07 1.5	07996e+08	1.564768e+07	1.340043e+07	1.782839e+07	1.364024e+07	6.051296e+06	
2016-12	30 1.780216e	+07 1.2	56468e+08	1.156067e+07	1.247502e+07	1.959575e+07	1.222803e+07	6.117870e+06	
2016-12	31 1.675861e	+07 1.2	38632e+08	1.107802e+07	1.150493e+07	2.460054e+07	1.338433e+07	6.305259e+06	
550 rows	× 7 columns								
trial_5.columns=['de', 'en', 'es', 'fr', 'ja', 'ru', 'zh']									
<pre>trial_5.reset_index(inplace=True)</pre>									
trial_5.columns=['date','de', 'en', 'es', 'fr', 'ja', 'ru', 'zh']									
<pre>trial_5.set_index('date',inplace=True)</pre>									
trial_5									

```
Adease.ipynb - Colaboratory
                          de
                                                     es
                                                                  fr
                                                                               jа
                                                                                                          zh
                                        en
                                                                                             ru
           date
      2015-07-01 1.507832e+07 9.415409e+07 1.618992e+07 9.232359e+06 1.549925e+07 1.170350e+07 5.159275e+06
      2015-07-02 1.489782e+07 9.387991e+07 1.551257e+07 9.286675e+06 1.725639e+07 1.186729e+07 5.165336e+06
      2015-07-03 1.437188e+07 8.960875e+07 1.433925e+07 8.959763e+06 1.594099e+07 1.116309e+07 5.137253e+06
      2015-07-04 1 333821e+07 9 290429e+07 1 351702e+07 9 521535e+06 1 909167e+07 1 063287e+07 5 176900e+06
      2015-07-05 1.521015e+07 9.563976e+07 1.462066e+07 9.362161e+06 1.846234e+07 1.117435e+07 5.454677e+06
      2016-12-27 2.032261e+07 1.458706e+08 1.594582e+07 1.528222e+07 1.626896e+07 1.520168e+07 6.487991e+06
      2016-12-28 1.934974e+07 1.415205e+08 1.657789e+07 1.378210e+07 1.629641e+07 1.416161e+07 6.522969e+06
      2016-12-29 1 864423e+07 1 507996e+08 1 564768e+07 1 340043e+07 1 782839e+07 1 364024e+07 6 051296e+06
      2016-12-30 1.780216e+07 1.256468e+08 1.156067e+07 1.247502e+07 1.959575e+07 1.222803e+07 6.117870e+06
      2016-12-31 1 675861e+07 1 238632e+08 1 107802e+07 1 150493e+07 2 460054e+07 1 338433e+07 6 305259e+06
trial_5.de
     date
     2015-07-01
                  1.507832e+07
     2015-07-02
                  1.489782e+07
     2015-07-03
                  1,437188e+07
     2015-07-04
                  1.333821e+07
     2015-07-05
                  1.521015e+07
     2016-12-27
                  2.032261e+07
     2016-12-28
                 1.934974e+07
     2016-12-29
                  1.864423e+07
     2016-12-30
                  1.780216e+07
     2016-12-31
                  1.675861e+07
     Name: de, Length: 550, dtype: float64
trial_5.index=pd.to_datetime(trial_5.index)
trial 5
                                                                                                               1
                                        en
                                                     es
           date
      2015-07-01 1.507832e+07 9.415409e+07 1.618992e+07 9.232359e+06 1.549925e+07 1.170350e+07 5.159275e+06
      2015-07-02 1.489782e+07 9.387991e+07 1.551257e+07 9.286675e+06 1.725639e+07 1.186729e+07 5.165336e+06
      2015-07-03 1.437188e+07 8.960875e+07 1.433925e+07 8.959763e+06 1.594099e+07 1.116309e+07 5.137253e+06
      2015-07-04 1.333821e+07 9.290429e+07 1.351702e+07 9.521535e+06 1.909167e+07 1.063287e+07 5.176900e+06
      2015-07-05 1.521015e+07 9.563976e+07 1.462066e+07 9.362161e+06 1.846234e+07 1.117435e+07 5.454677e+06
      2016-12-27 2.032261e+07 1.458706e+08 1.594582e+07 1.528222e+07 1.626896e+07 1.520168e+07 6.487991e+06
      2016-12-28 1.934974e+07 1.415205e+08 1.657789e+07 1.378210e+07 1.629641e+07 1.416161e+07 6.522969e+06
      2016-12-29 1.864423e+07 1.507996e+08 1.564768e+07 1.340043e+07 1.782839e+07 1.364024e+07 6.051296e+06
      2016-12-30 1.780216e+07 1.256468e+08 1.156067e+07 1.247502e+07 1.959575e+07 1.222803e+07 6.117870e+06
      2016-12-31 1.675861e+07 1.238632e+08 1.107802e+07 1.150493e+07 2.460054e+07 1.338433e+07 6.305259e+06
     550 rows × 7 columns
import statsmodels.api as sm
```

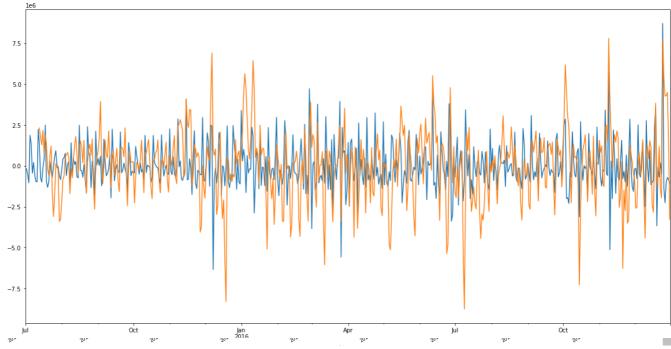
```
model = sm.tsa.seasonal_decompose(trial_5['de'], model='additive')
plt.rcParams['figure.figsize'] = (20, 10)
# dickey fuller test and decomposition
for i in trial_5.columns:
 model = sm.tsa.seasonal_decompose(trial_5[i], model='additive')
  model.plot()
 model.resid.plot()
```

```
plt.show()
pvalue = sm.tsa.stattools.adfuller(model.resid.dropna())[1]
if pvalue <= 0.05:
    print(f'the time series residual for the language {i} is stationary')
else:
    print(f'the time series residual for the language {i} is notstationary')</pre>
```



```
trial_5['de'].diff().plot()
trial_5['de'].diff(12).plot()
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f17e21f51c0>
```



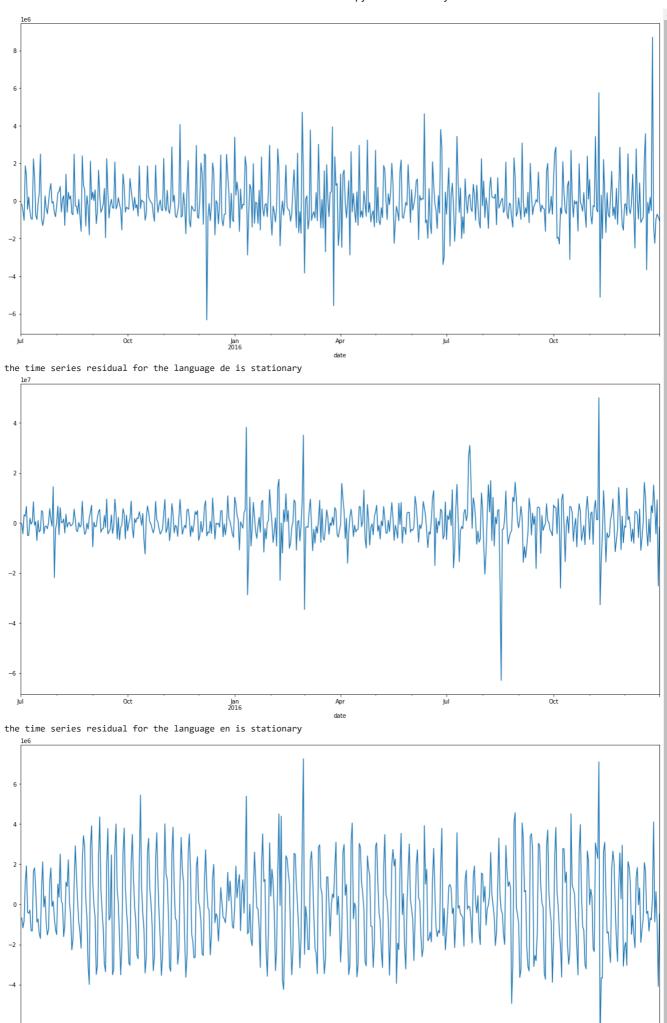
trial_5['de'].diff(),trial_5['de'].diff().diff()

```
(date
2015-07-01
                       NaN
2015-07-02
             -1.805018e+05
2015-07-03
             -5.259441e+05
2015-07-04
             -1.033667e+06
2015-07-05
              1.871935e+06
             -2.251612e+06
2016-12-27
2016-12-28
            -9.728757e+05
2016-12-29
             -7.055054e+05
             -8.420762e+05
2016-12-30
2016-12-31
            -1.043543e+06
Name: de, Length: 550, dtype: float64, date
2015-07-01
2015-07-02
                       NaN
2015-07-03
            -3.454423e+05
2015-07-04
             -5.077229e+05
2015-07-05
              2.905602e+06
             -8.364752e+05
2016-12-27
2016-12-28
             1.278736e+06
2016-12-29
              2.673703e+05
2016-12-30
             -1.365708e+05
2016-12-31
             -2.014666e+05
Name: de, Length: 550, dtype: float64)
   1.2
```

```
for i in trial_5.columns:
    trial_5[i].diff().plot()
    plt.show()
    pvalue = sm.tsa.stattools.adfuller(trial_5[i].diff().diff(1).dropna())[1]
    if pvalue <= 0.05:</pre>
```

```
\label{eq:print}  \text{print(f'the time series residual for the language \{i\} is stationary')} \\ \text{else:} \\ \text{print(f'the time series residual for the language \{i\} is not stationary')} \\
```

https://colab.research.google.com/drive/1n6uYbVTgZjkOn3jHV078ezkV6x3cHiA8#scrollTo=k9sHSUcXbMfq&printMode=true

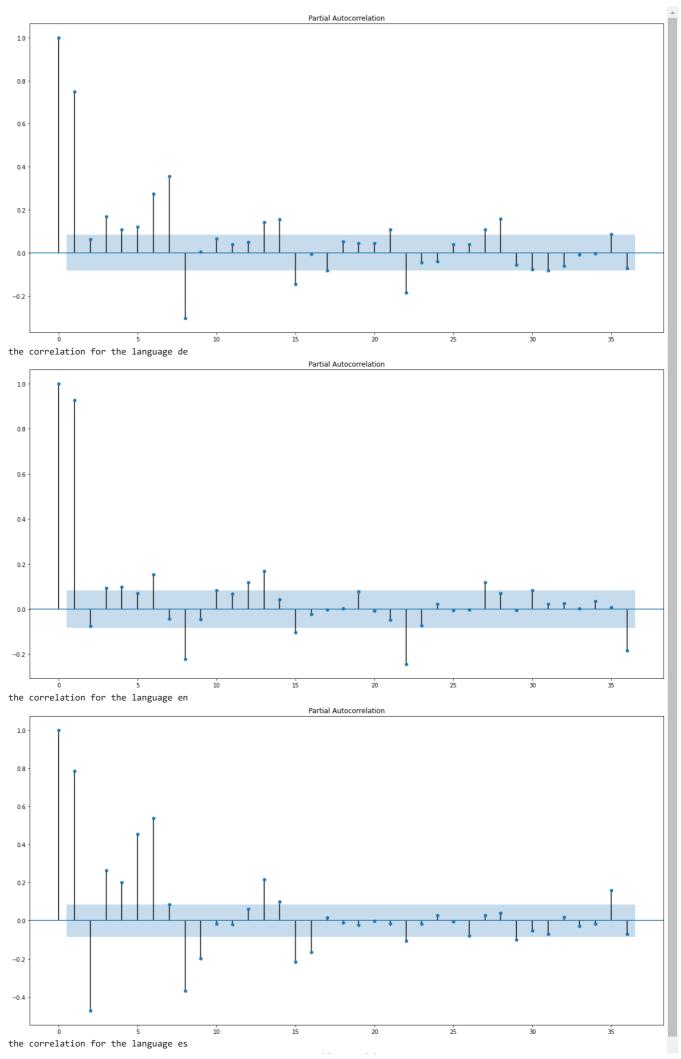


With both decomposition and differentiation we find that the time series with respect to data is stationary

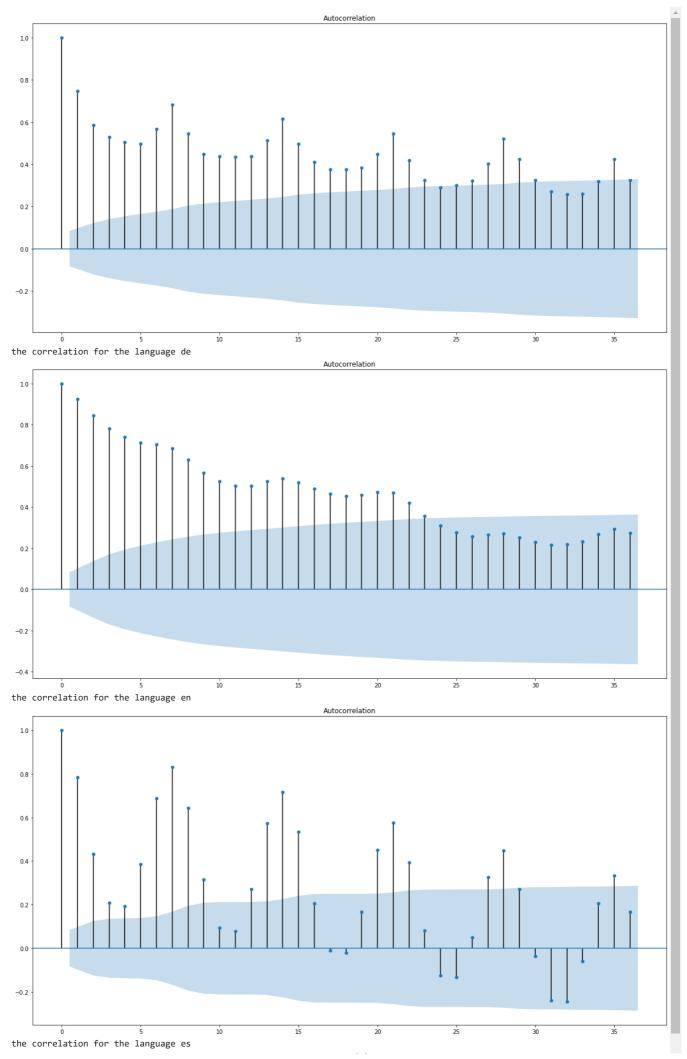
Creating model training and forecasting with ARIMA, SARIMAX

from statsmodels.graphics.tsaplots import plot_acf, plot_pacf

for i in trial_5.columns:
 plot_pacf(trial_5[i],lags=36)
 plt.show()
 print(f'the correlation for the language {i}')



Partial Autocorrelation



Autocorrelation

```
# the corresponding q values for each language is 23,22,10,30,22,14,30
       1
                                                                                                                     # for arima d is required consider d values
       1
            1 1
                                                                                                                     from statsmodels.tsa.statespace.sarimax import SARIMAX
            1 1 1 1 1
      0.6
                                                                                                                     p=[2,2,7,4,3,4,2]
            \begin{smallmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{smallmatrix}
       q=[23,22,10,30,22,14,30]
            1
                                                                                                                     np.argmax(p)
    2
            . . L
trial_5.shape[1]
    7
from sklearn.metrics import (
   mean_squared_error as mse,
   mean_absolute_error as mae,
   mean_absolute_percentage_error as mape
)
# Creating a function to print values of all these metrics.
def performance(actual, predicted):
   # print('MAE :', round(mae(actual, predicted), 3))
   # print('RMSE :', round(mse(actual, predicted)**0.5, 3))
   return round(mape(actual, predicted), 3)
      ... 1 1
                                                                                                                     for i in range(trial_5.shape[1]):
 train_x = pd.DataFrame(trial_5.loc[trial_5.index < trial_5.index[-55]].copy().iloc[:,i])</pre>
 test_x = pd.DataFrame(trial_5.loc[trial_5.index >= trial_5.index[-55]].copy().iloc[:,i])
 array=[]
 for d in range(1,3):
   model = SARIMAX(train_x.iloc[:,0], order=(p[i], d, q[i]))
   model = model.fit(disp=False)
   test_x.loc[:,'pred'] = model.forecast(steps=55)
   array.append(performance(test_x.iloc[:,0], test_x.loc[:,'pred']))
 best d=np.argmax(array)
 model = SARIMAX(train_x.iloc[:,0], order=(p[i], best_d, q[i]))
  model = model.fit(disp=False)
 test_x.loc[:,'pred'] = model.forecast(steps=55)
 test_x.iloc[:,0].plot(style='-o')
 test_x.loc[:,'pred'].plot(style='-o')
 plt.show()
  a=performance(test_x.iloc[:,0], test_x.loc[:,'pred'])
 print(f'The performance is {a} for the language {trial_5.columns[i]}')
```

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

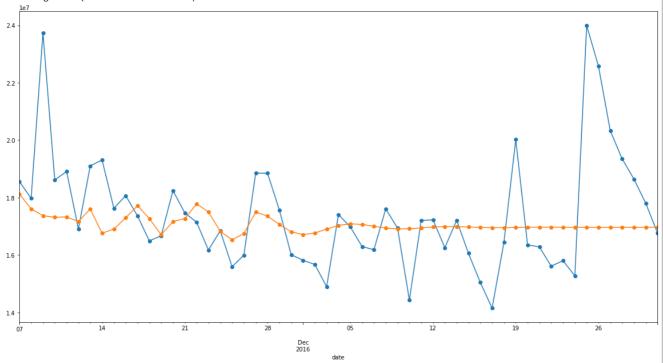
/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/statespace/sarimax.py:978: UserWarning: Non-invertible starting MA parameters found.'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "



The performance is 0.072 for the language de

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

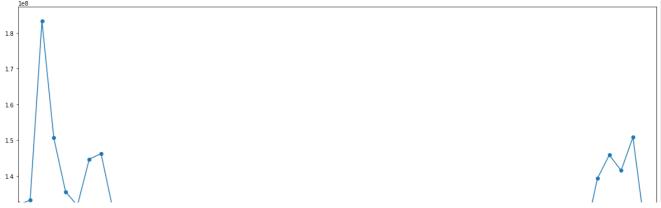
/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/statespace/sarimax.py:978: UserWarning: Non-invertible starting MA paramet warn('Non-invertible starting MA parameters found.'

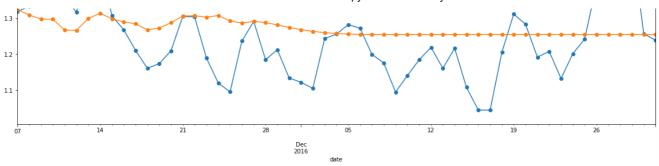
/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "





The performance is 0.076 for the language en

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

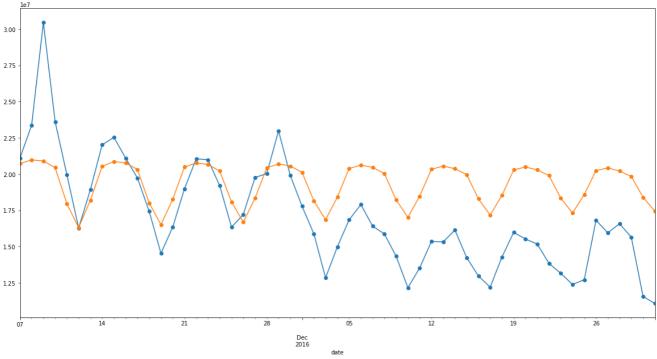
/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "



The performance is 0.211 for the language es

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/statespace/sarimax.py:978: UserWarning: Non-invertible starting MA parameters found.'

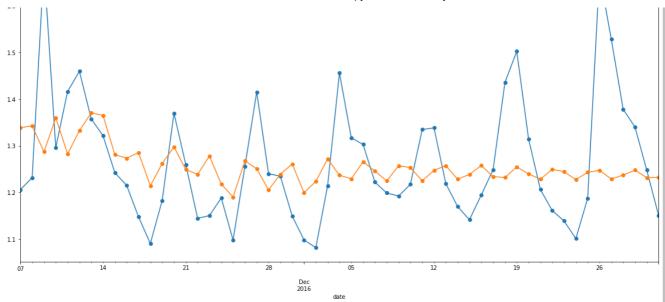
/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "





The performance is 0.075 for the language fr

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

warnings.warn('No frequency information was'
warnings.warn('No frequency information was')

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

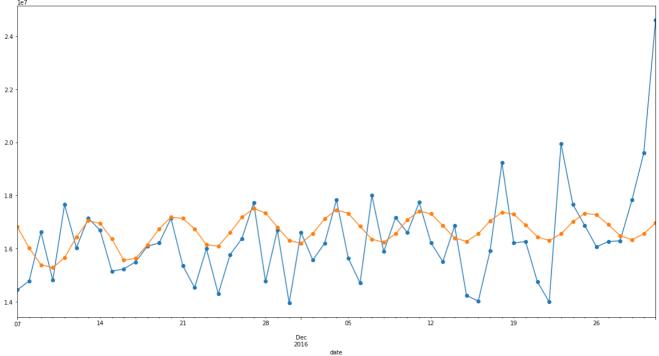
/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/statespace/sarimax.py:978: UserWarning: Non-invertible starting MA paramet warn('Non-invertible starting MA parameters found.'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "



The performance is 0.075 for the language ja

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

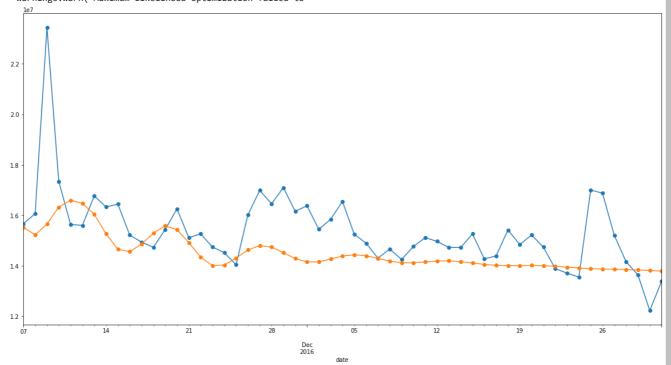
/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed

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/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "



The performance is 0.065 for the language ru

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings.warn('No frequency information was'

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/statespace/sarimax.py:978: UserWarning: Non-invertible starting MA parameters found.'

/usr/local/lib/python3.8/dist-packages/statsmodels/base/model.py:566: ConvergenceWarning: Maximum Likelihood optimization failed warnings.warn("Maximum Likelihood optimization failed to "

/usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided warnings warn('No frequency information was'

trial_5.index

/usr/local/lib/nython3 %/dist-packages/statsmodels/base/model ny:566: ConvergenceWarning: Maximum Likelihood ontimization failed exog_1.set_index(trial_5.index,inplace=True)

9.0 -

exog_1

	_	+++
	Exog	1
date		
2015-07-01	0	
2015-07-02	0	
2015-07-03	0	
2015-07-04	0	
2015-07-05	0	
2016-12-27	1	
2016-12-28	1	
2016-12-29	1	
2016-12-30	0	
2016-12-31	0	
550 rows × 1	columns	

```
trial_5=trial_5.join(exog_1)
```

trial_5

```
zh Exog
                                                                                                                      1
                          de
                                        en
                                                     es
           date
      2015-07-01 1.507832e+07 9.415409e+07 1.618992e+07 9.232359e+06 1.549925e+07 1.170350e+07 5.159275e+06
                                                                                                                  0
      2015-07-02 1.489782e+07 9.387991e+07 1.551257e+07 9.286675e+06 1.725639e+07 1.186729e+07 5.165336e+06
                                                                                                                  O
      2015-07-03 1.437188e+07 8.960875e+07 1.433925e+07 8.959763e+06 1.594099e+07 1.116309e+07 5.137253e+06
      2015-07-04 1.333821e+07 9.290429e+07 1.351702e+07 9.521535e+06 1.909167e+07 1.063287e+07 5.176900e+06
      2015-07-05 1 521015e+07 9 563976e+07 1 462066e+07 9 362161e+06 1 846234e+07 1 117435e+07 5 454677e+06
                                                                                                                  0
      2016-12-27 2.032261e+07 1.458706e+08 1.594582e+07 1.528222e+07 1.626896e+07 1.520168e+07 6.487991e+06
                                                                                                                  1
      2016-12-28 1.934974e+07 1.415205e+08 1.657789e+07 1.378210e+07 1.629641e+07 1.416161e+07 6.522969e+06
                                                                                                                  1
      2016-12-29 1.864423e+07 1.507996e+08 1.564768e+07 1.340043e+07 1.782839e+07 1.364024e+07 6.051296e+06
      2016-12-30 1.780216e+07 1.256468e+08 1.156067e+07 1.247502e+07 1.959575e+07 1.222803e+07 6.117870e+06
                                                                                                                  0
      2016-12-31 1.675861e+07 1.238632e+08 1.107802e+07 1.150493e+07 2.460054e+07 1.338433e+07 6.305259e+06
     550 rows × 8 columns
# sarimax only for english
train_x = pd.DataFrame(trial_5.loc[trial_5.index < trial_5.index[-55]].copy().iloc[:,[1,-1]])</pre>
test_x = pd.DataFrame(trial_5.loc[trial_5.index >= trial_5.index[-55]].copy().iloc[:,[1,-1]])
model = SARIMAX(train_x['en'],exog=train_x['Exog'],order=(1,1,1),seasonal_order=(1,1,1,7),enforce_invertibility=False)
     /usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided,
      warnings.warn('No frequency information was
     /usr/local/lib/python3.8/dist-packages/statsmodels/tsa/base/tsa_model.py:524: ValueWarning: No frequency information was provided,
      warnings.warn('No frequency information was'
results = model.fit(disp=False)
exog forecast = test x[['Exog']]
predictions = results.predict(start=train_x.shape[0], end=train_x.shape[0]+test_x.shape[0]-1, exog=exog_forecast).rename('Predictions')
performance(test_x['en'], predictions)
     0.098
```

Forecasting with prophet

```
!pip install pystan~=2.14
!pip install fbprophet
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Collecting pystan~=2.14
       Downloading pystan-2.19.1.1-cp38-cp38-manylinux1_x86_64.whl (62.6 MB)
                                                      - 62.6/62.6 MB 15.6 MB/s eta 0:00:00
     Requirement already satisfied: numpy>=1.7 in /usr/local/lib/python3.8/dist-packages (from pystan~=2.14) (1.21.6)
     Requirement already satisfied: Cython!=0.25.1,>=0.22 in /usr/local/lib/python3.8/dist-packages (from pystan~=2.14) (0.29.33)
     Installing collected packages: pystan
       Attempting uninstall: pystan
         Found existing installation: pystan 3.3.0
         Uninstalling pystan-3.3.0:
           Successfully uninstalled pystan-3.3.0
     Successfully installed pystan-2.19.1.1
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Collecting fbprophet
       Downloading fbprophet-0.7.1.tar.gz (64 kB)
                                                      - 64.0/64.0 KB 2.5 MB/s eta 0:00:00
       Preparing metadata (setup.py) ... done
     Requirement already satisfied: Cython>=0.22 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (0.29.33)
     Collecting cmdstanpy==0.9.5
       Downloading cmdstanpy-0.9.5-py3-none-any.whl (37 kB)
     Requirement already satisfied: pystan>=2.14 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (2.19.1.1) Requirement already satisfied: numpy>=1.15.4 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (1.21.6)
     Requirement already satisfied: pandas>=1.0.4 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (1.3.5)
```

```
Requirement already satisfied: matplotlib>=2.0.0 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (3.2.2)
     Requirement already satisfied: LunarCalendar>=0.0.9 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (0.0.9)
     Requirement already satisfied: convertdate>=2.1.2 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (2.4.0)
     Requirement already satisfied: holidays>=0.10.2 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (0.19)
     Collecting setuptools-git>=1.2
       Downloading setuptools_git-1.2-py2.py3-none-any.whl (10 kB)
     Requirement already satisfied: python-dateutil>=2.8.0 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (2.8.2)
     Requirement already satisfied: tqdm>=4.36.1 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (4.64.1)
     Requirement already satisfied: pymeeus<=1,>=0.3.13 in /usr/local/lib/python3.8/dist-packages (from convertdate>=2.1.2->fbprophet)
     Requirement already satisfied: korean-lunar-calendar in /usr/local/lib/python3.8/dist-packages (from holidays>=0.10.2->fbprophet) (
     Requirement already satisfied: hijri-converter in /usr/local/lib/python3.8/dist-packages (from holidays>=0.10.2->fbprophet) (2.2.4)
     Requirement already satisfied: ephem>=3.7.5.3 in /usr/local/lib/python3.8/dist-packages (from LunarCalendar>=0.0.9->fbprophet) (4.1
     Requirement already satisfied: pytz in /usr/local/lib/python3.8/dist-packages (from LunarCalendar>=0.0.9->fbprophet) (2022.7.1)
     Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.8/dist-packages (from matplotlib)
     Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.8/dist-packages (from matplotlib>=2.0.0->fbprophet) (1.4
     Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.8/dist-packages (from matplotlib>=2.0.0->fbprophet) (0.11.0)
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.8/dist-packages (from python-dateutil>=2.8.0->fbprophet) (1.15.0)
     Building wheels for collected packages: fbprophet
       Building wheel for fbprophet (setup.py) ... done
       Created wheel for fbprophet: filename=fbprophet-0.7.1-py3-none-any.whl size=9537455 sha256=08785b29ec967aceda6e84114d239b2e87e8ac
       Stored in directory: /root/.cache/pip/wheels/d0/d2/ae/c579b7fd160999d35908f3cb8ebcad7ef64ecaca7b78e4c3c8
     Successfully built fbprophet
     Installing collected packages: setuptools-git, cmdstanpy, fbprophet
       Attempting uninstall: cmdstanpy
         Found existing installation: cmdstanpy 1.1.0
         Uninstalling cmdstanpy-1.1.0:
           Successfully uninstalled cmdstanpy-1.1.0
     ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the sc
     prophet 1.1.2 requires cmdstanpy>=1.0.4, but you have cmdstanpy 0.9.5 which is incompatible. Successfully installed cmdstanpy-0.9.5 fbprophet-0.7.1 setuptools-git-1.2
exog=exog_1['Exog'].to_numpy()
trial_6= trial_5.iloc[:,[1]].copy().reset_index()
trial_6
                                      10:
                date
                                en
          2015-07-01 9.415409e+07
           2015-07-02 9.387991e+07
           2015-07-03 8.960875e+07
       2
           2015-07-04 9.290429e+07
       3
       4
           2015-07-05 9 563976e+07
      545 2016-12-27 1.458706e+08
      546 2016-12-28 1.415205e+08
      547 2016-12-29 1.507996e+08
      548 2016-12-30 1.256468e+08
      549 2016-12-31 1.238632e+08
     550 rows × 2 columns
trial_6.columns = [['ds', 'y']]
# trial_6=df.copy()
trial 6['exog'] = exog
trial_6.columns = ['ds', 'y', 'exog']
# trial_6.head()
from prophet import Prophet
model=Prophet(weekly_seasonality=True)
model.add_regressor('exog')
model.fit(trial_6[:-55])
forecast = model.predict(trial_6)
fig = model.plot(forecast)
```