```
In [1]: import pandas as pd
        import re
        from emoji import UNICODE EMOJI
        from textblob import TextBlob
        import altair as alt
        import numpy as np
        from collections import Counter
        import string
        import nltk
        nltk.download('vader lexicon')
        nltk.download('brown')
        nltk.download('punkt')
        nltk.download('stopwords')
        from nltk.tokenize import sent tokenize, word tokenize
        from nltk.corpus import stopwords
        import matplotlib.pyplot as plt
        %matplotlib inline
        [nltk data] Downloading package vader lexicon to
        [nltk data]
                        /home/jovyan/nltk data...
        [nltk data] Package vader lexicon is already up-to-date!
        [nltk data] Downloading package brown to /home/jovyan/nltk data...
```

```
[nltk_data] Package vader_lexicon is already up-to-date!
[nltk_data] Downloading package brown to /home/jovyan/nltk_data...
[nltk_data] Package brown is already up-to-date!
[nltk_data] Downloading package punkt to /home/jovyan/nltk_data...
[nltk_data] Package punkt is already up-to-date!
[nltk_data] Downloading package stopwords to /home/jovyan/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

The data cleaning/manipulation technique/functions

```
In [2]: | def extract tags(text):
            return re.findall("#([a-zA-Z0-9]{1,50})", text)
        def extract emoji(text):
            return [ch for ch in text if ch in UNICODE EMOJI['en']]
        def clean_tweet(txt):
            temp = re.sub("@[A-Za-z0-9_]+","", txt)
            temp1 = re.sub("#[A-Za-z0-9_]+","", temp)
            temp2 = re.sub(r"http\S+", "", temp1)
            result=''.join(i for i in temp2.lower() if (i.isalpha() or i==' '))
             return result
        def word list(tweet):
            lst = word_tokenize(tweet)
            lst1 = []
            stops = list(stopwords.words('english'))
            for w in 1st:
                 if w not in stops:
                     lst1.append(w)
             return 1st1
        def sentiment(tweet):
             blob = TextBlob(tweet)
            return blob.sentiment.polarity
        def get_date(date):
            return date[:10]
        def get hour(date):
            return date[11:13]
        def get_10min(date):
            return date[14]+'0'
        def get_min(date):
            return date[14:16]
```

```
def firm pos(score):
    if score \geq= 0.7:
        return 1
    else: return 0
def pos(score):
    if (score >= 0.25) & (score < 0.7):</pre>
        return 1
    else: return 0
def neutral(score):
    if (score >= -0.25) & (score < 0.25):
        return 1
    else: return 0
def neg(score):
    if (score > -0.7) & (score < -0.25):
        return 1
    else: return 0
def firm_neg(score):
    if score <= -0.7:
        return 1
    else: return 0
```

Import data, check duplicate and missing/incomplete data. Remove if exists

Apply the data score, and spe	a cleaning/manipul ecific date/hour/mi	lation techniques in data.	on the data, we	e now have the	used words, tag	ıs, emojis, sentim

```
In [6]: df['tags']= df.apply(lambda row: extract_tags(row['text']), axis=1)
    df['emojis']= df.apply(lambda row: extract_emoji(row['text']), axis=1)
    df['clean_text']= df.apply(lambda row: clean_tweet(row['text']), axis=1)
    df['words']= df.apply(lambda row: word_list(row['clean_text']), axis=1)
    df['sentiment_score']= df.apply(lambda row: sentiment(row['clean_text']), axis=1)
    df['day']= df.apply(lambda row: get_date(row['date']), axis=1)
    df['hour']= df.apply(lambda row: get_l0min(row['date']), axis=1)
    df['l0min']= df.apply(lambda row: get_min(row['date']), axis=1)
    df['min']= df.apply(lambda row: get_min(row['date']), axis=1)
    df['POS']= df.apply(lambda row: firm_pos(row['sentiment_score']), axis=1)
    df['neu']= df.apply(lambda row: neutral(row['sentiment_score']), axis=1)
    df['neu']= df.apply(lambda row: neg(row['sentiment_score']), axis=1)
    df['neg']= df.apply(lambda row: firm_neg(row['sentiment_score']), axis=1)
    df['NEG']= df.apply(lambda row: firm_neg(row['sentiment_score']), axis=1)
```

Out[6]:

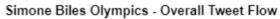
•	id	date	text	tags	emojis	clean_text	words	sentiment_score	day	hour	10min	min	POS	pos
_	0 1418360000888647681	2021-07-22 23:59:10+00:00	the olympics r literally like tmrw??? anyways		0	the olympics r literally like tmrw anyways can	[olympics, r, literally, like, tmrw, anyways,		2021- 07-22	23	50	59	0	0
	1 1418359846253047810	2021-07-22 23:58:33+00:00	omg it's the goat #SimoneBiles LOOK AT THE GOAT	[SimoneBiles]	0	omg its the goat look at the goat	[omg, goat, look, goat]		2021- 07-22		50	58	0	0
	2 1418359748500418573	2021-07-22 23:58:10+00:00		п	0	simone biles is a gymnast and an entertainerim	[simone, biles, gymnast, entertainerim, glad,	0.166667	2021- 07-22	23	50	58	0	0
	3 1418359695681728515	2021-07-22 23:57:57+00:00		[SimoneBiles]	0	just wanted to see the emoji	[wanted, see, emoji]	0.000000	2021- 07-22	23	50	57	0	0
	4 1418359661028331526	2021-07-22 23:57:49+00:00	God, I'm not worthy enough to share a planet w	0	0	god im not worthy enough to share a planet wit		-0.083333	2021- 07-22	23	50	57	0	0
4														

See the overall flow of tweet & sentiment

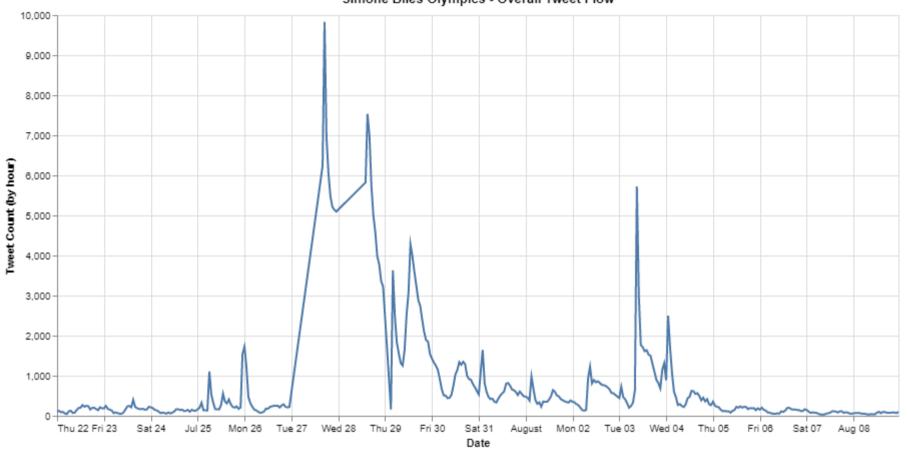
Out[7]:

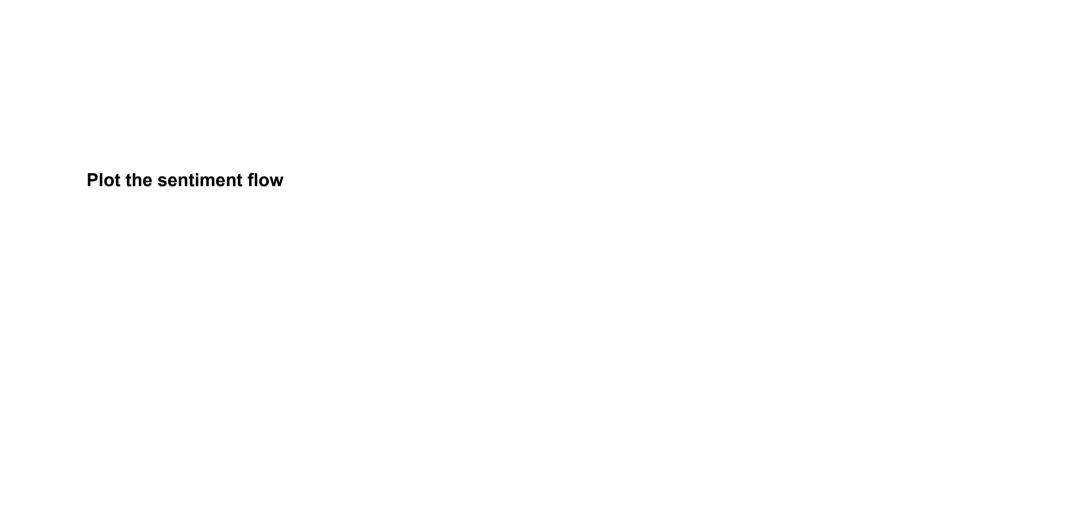
	day	hour	sum	size	date	6hr_sum	6hr_count	6hr_avg
0	2021-07-22	00	28.321041	106.0	2021-07-22 00:00	28.321041	106.0	0.267180
1	2021-07-22	01	27.016179	125.0	2021-07-22 01:00	55.337220	231.0	0.239555
2	2021-07-22	02	18.577410	88.0	2021-07-22 02:00	73.914630	319.0	0.231707
3	2021-07-22	03	16.880703	97.0	2021-07-22 03:00	90.795333	416.0	0.218258
4	2021-07-22	04	12.141448	57.0	2021-07-22 04:00	102.936781	473.0	0.217625







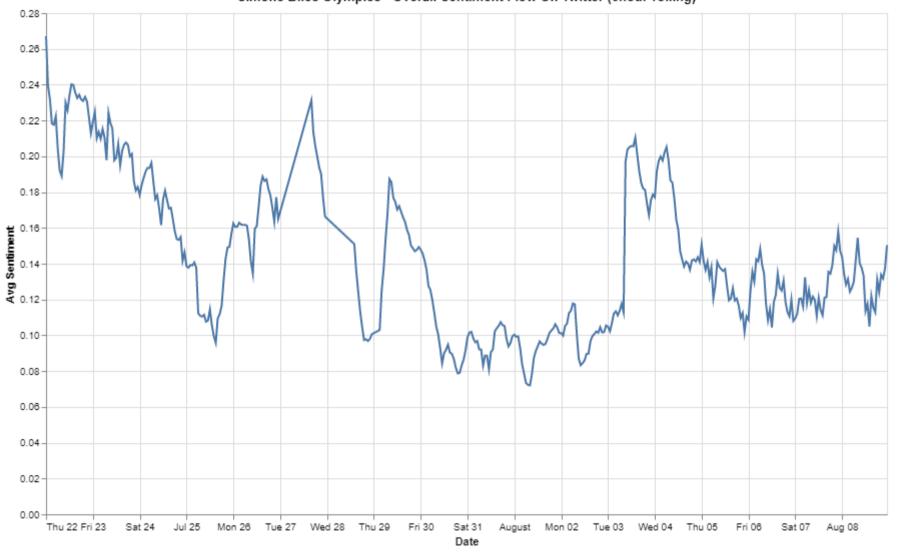












```
In [10]: flow = df.groupby(['day', 'hour']).mean()[['POS', 'pos', 'neu', 'neg', 'NEG']]
         flow = flow.reset index()
         flow['date'] = flow['day'] + ' ' + flow['hour'] + ':00'
         flow[['POSITIVE', 'positive', 'neutral', 'negative', 'NEGATIVE']] = flow.rolling(window=6,min periods=1).mean()[['POS','pos','neu']
         flow1 = pd.DataFrame()
         dates = []
         values = []
         labels = []
         for i in ['POSITIVE', 'positive', 'neutral', 'negative', 'NEGATIVE']:
             lst = []
             lst1 = []
             lst2 = list(flow.date.values)
             for j in range(len(flow)):
                  lst.append(i)
                  lst1.append(flow[i][j])
             dates += 1st2
             labels += lst
             values += lst1
         flow1['date'] = pd.Series(dates)
         flow1['sentiment label'] = pd.Series(labels)
         flow1['percentage'] = pd.Series(values)
         flow1.head(10)
```

Out[10]:

			1
0	2021-07-22 00:00	POSITIVE	0.141509
1	2021-07-22 01:00	POSITIVE	0.102755
2	2021-07-22 02:00	POSITIVE	0.087443
3	2021-07-22 03:00	POSITIVE	0.083623
4	2021-07-22 04:00	POSITIVE	0.084442
5	2021-07-22 05:00	POSITIVE	0.093096
6	2021-07-22 06:00	POSITIVE	0.076951
7	2021-07-22 07:00	POSITIVE	0.078096

date sentiment label percentage

```
date sentiment_label percentage
8 2021-07-22 08:00
                         POSITIVE
                                     0.078016
                         POSITIVE
9 2021-07-22 09:00
                                      0.076953
```

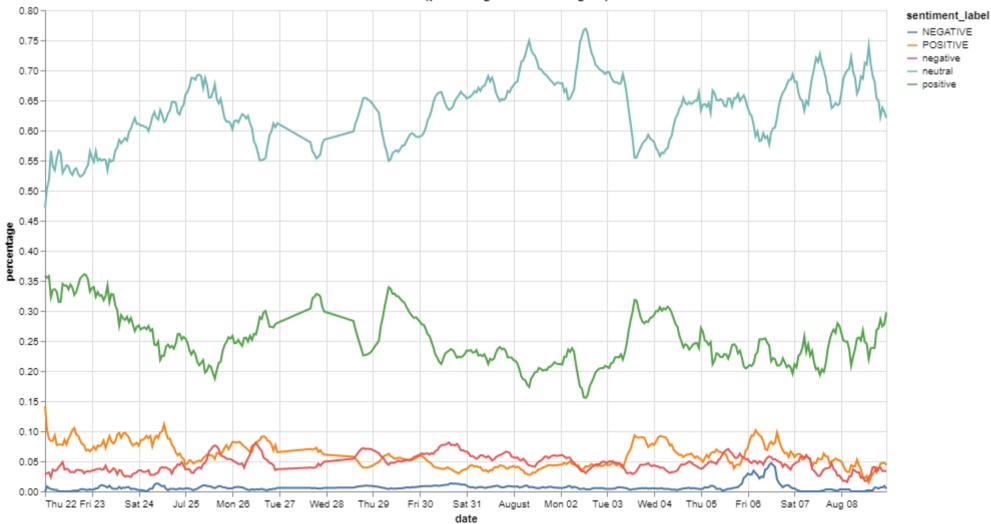
```
In [11]:
         alt.Chart(flow1).mark_line().encode(
             x='date:T',
             y='percentage:Q',
             color='sentiment label:N'
         ).properties(width=840,height=480,title='sentiments flow (percentage of each 5 degree)')
```

Out[11]:

sentiments flow (percentage of each 5 degree)

POSITIVE

neutral



Emoji/Tags

```
In [12]: # this return the top 50 most common items in the columns (emoji/tag/word)

def top_item(data,label):
    lst = []
    for i in data[label]:
        lst += i

    C = Counter(lst)
    top50 = C.most_common(50)
    count_df = pd.DataFrame(top50,columns = [label,'count'])

    return count_df
```

```
In [13]: c= top_item(df,'tags')
c
# the top 50 most frequently used tags within the 'simone biles' tweets during the 2021 olympics
```

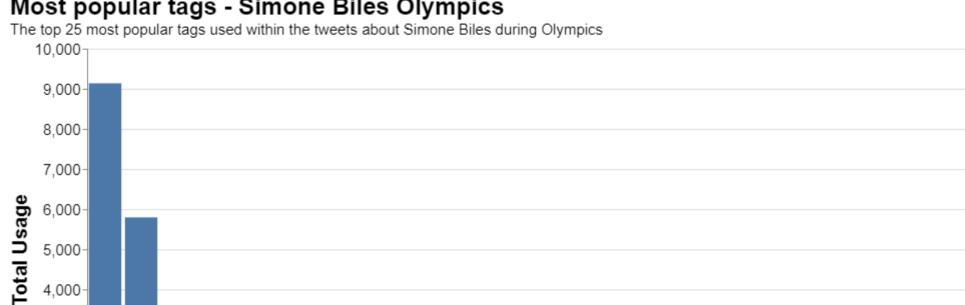
Out[13]:

	tags	count
0	SimoneBiles	26354
1	Olympics	9137
2	Tokyo2020	5795
3	TokyoOlympics	3054
4	Simone	2835
5	MentalHealthMatters	2327
6	GOAT	2214
7	mentalhealth	2030
8	TeamUSA	1900
9	OlympicGames	1841
10	simonebiles	1631
11	gymnastics	1279
12	ArtisticGymnastics	1204
13	Olympics2021	1157
14	USA	822
15	Gymnastics	808
16	USAGymnastics	724
17	olympics	712
18	NaomiOsaka	705
19	MentalHealth	592
20	simone	518
21	Biles	476
22	Olympics2020	426
23	MentalHealthAwareness	398
24	TokyoOlympics2020	380

	tags	count
25	FoxNews	300
26	1	295
27	tokyo2020	284
28	Olympic	284
29	Mentalhealth	281
30	goat	269
31	usa	248
32	selfcare	245
33	PiersMorgan	242
34	SuniLee	236
35	SmartNews	233
36	Tokyo	224
37	SimonBiles	220
38	SimoneStrong	213
39	BlackGirlMagic	206
40	teamusa	199
41	NewsBreak	190
42	news	184
43	simonebilesgoat	182
44	SIMONEBILES	179
45	BREAKING	177
46	mentalhealthmatters	177
47	TokyoOlympics2021	171
48	sports	171
49	Tokyo2021	168

Out[14]:

Most popular tags - Simone Biles Olympics



Artistic Symnastics

gynnastics

Olympics 2021



USA Gyrina stics

Gymnastics

Nagrijo saka

OWNIDICS

Mentallealth

simone

Mental Health Awareness

Takyo Olympics 2020

FOXYEWS

4,000

3,000-

2,000

1,000-

0

MenalHealthMatters

OwnpicSames

TeamUSA

simonediles

nentalhealth

CORT

TOKYODYMPICS

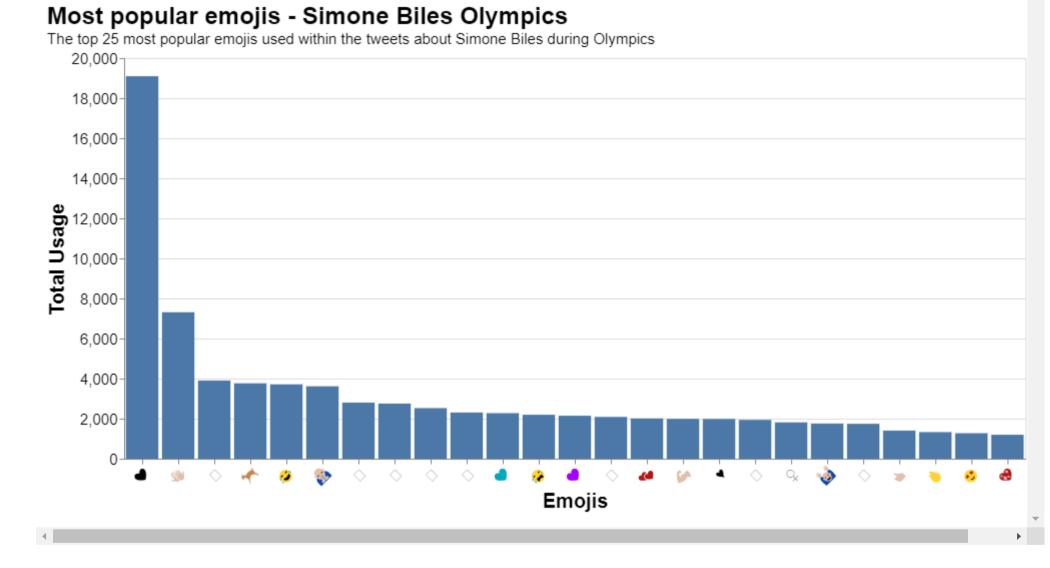
LOHADDO

```
In [15]: count = top_item(df, 'emojis')
         count
```

Out[15]:			
		emojis	count
	0	•	19096
	1	*	7312
	2		3903
	3	*	3764
	4	8	3712
	5		3616
	6		2802
	7		2753
	8		2526
	9		2307
	10	₩	2275
	11	0	2195
	12	4	2146
	13		2088
	14	•	2012
	15	4	1992
	16	•	1985
	17		1939
	18	\$	1814
	19	•	1757
	20		1743
	21	•	1404
	22	٠	1328
	23	•	1273
	24	⊗	1197

	emojis	count
25		1184
26	<u>100</u>	1146
27		1122
28	8	1106
29		1045
30		1003
31		991
32		975
33		886
34	+ \$	867
35		859
36		838
37	•	782
38		727
39	ⓒ	711
40	췐	652
41	<i>\(\mathref{m}\)</i>	647
42		638
43		621
44		591
45	\odot	549
46	*	549
47		509
48	W	506
49		447

Out[16]:



Specifically target the period after the withdrawl, see how Twitter react

```
In [17]: df['Date'] = pd.to_datetime(df['date'])
    mask = (df['Date'] > '2021-07-27 00:00') & (df['Date'] < '2021-07-28 23:59')
    drawl = df.loc[mask].sort_values('Date')
    drawl = drawl.reset_index()
    drawl.drop(columns=['index','Date'],inplace=True)

drawl.head()</pre>
```

Out[17]:

•		id	date	text	tags	emojis	clean_text	words	sentiment_score	day	hour	10min	min	POS	pos	neu
-	0	1420059487524765703	2021-07-27 16:32:19+00:00	@NYCMayor @Simone_Biles Omg what a hero! Remem	[]	0	omg what a hero remember the days we used to	[omg, hero, remember, days, used, praise, real	-0.100000	2021- 07-27	16	30	32	0	0	1
	1	1420059488002920459	2021-07-27 16:32:20+00:00	The conversation around Simone Biles is exactl	[]	0	the conversation around simone biles is exactl	[conversation, around, simone, biles, exactly,	0.525000	2021- 07-27	16	30	32	0	1	0
	2	1420059487931617281	2021-07-27 16:32:20+00:00	@Simone_Biles @lourdesgnavarro ♥ you, Simone		[♥, ♥]	you simone and so proud of you	[simone, proud]	0.800000	2021- 07-27	16	30	32	1	0	0
	3	1420059491706433536	2021-07-27 16:32:20+00:00	@Simone_Biles @tonyposnanski You rock so hard!		0	you rock so hard	[rock, hard]	-0.291667	2021- 07-27	16	30	32	0	0	0
	4	1420059491484110848	2021-07-27 16:32:20+00:00	Logging in to say: good for Simone Biles.\nShe	0	0	logging in to say good for simone bilesshe has	[logging, say, good, simone, bilesshe, nothing	0.075000	2021- 07-27	16	30	32	0	0	1

```
In [18]: c = top_item(drawl,'tags')
c
```

Out[18]:

	tags	count
0	SimoneBiles	9193
1	Olympics	2427
2	MentalHealthMatters	1225
3	Tokyo2020	1079
4	GOAT	926
5	Simone	876
6	mentalhealth	746
7	TokyoOlympics	583
8	TeamUSA	506
9	OlympicGames	499
10	gymnastics	393
11	NaomiOsaka	383
12	simonebiles	379
13	Olympics2021	312
14	USAGymnastics	289
15	MentalHealth	244
16	USA	186
17	MentalHealthAwareness	184
18	olympics	162
19	ArtisticGymnastics	161
20	PiersMorgan	129
21	1	128
22	selfcare	119
23	simonebilesgoat	102
24	SimonBiles	93
25	Mentalhealth	92

	tags	count
26	goat	87
27	Olympics2020	86
28	usa	85
29	simone	80
30	mentalhealthmatters	78
31	Biles	75
32	SimoneStrong	73
33	Tokyo	73
34	Gymnastics	67
35	Tokyo2021	67
36	tokyo2020	62
37	Olympic	61
38	Respect	57
39	MeghanMarkle	57
40	ActiBlizzWalkout	54
41	BlackGirlMagic	49
42	olympics2021	46
43	womensgymnastics	44
44	January6thCommission	44
45	TokyoOlympics2020	43
46	edit	42
47	fancam	42
48	followtrick	41
49	kpop	41

Out[19]:

Most frequent tags - Simone Biles Withdrawl

The top 25 most frequently used tags within the tweets about Simone Biles after withdraw from Olympics

