EDOS Data Analysis

Reading the data

In []: import pandas as pd

```
pd.options.display.max_rows
          pd.set_option('display.max_colwidth', None)
          edos = pd.read_csv('../edos/data/edos_labelled_aggregated.csv')
          edos.set_index('rewire_id', inplace=True)
          edos
Out[]:
                                                              text label_sexist label_category
                                                                                                     label_vector split
                      rewire_id
                                    In Nigeria, if you rape a woman,
                                    the men rape you back! #NSFW
          sexism2022_english-
                                    "In Nigeria, if you rape a woman,
                                                                       not sexist
                                                                                            none
                                                                                                            none
                                                                                                                    dev
                          9609
                                     guys rape you back!" by [USER]
                                                              [URL]
           sexism2022_english-
                                           Then, she's a keeper. 😊
                                                                       not sexist
                                                                                            none
                                                                                                            none train
                         16993
                                      This is like the Metallica video
                                  where the poor mutilated bastard
          sexism2022_english-
                                    was saying "Please kill me" over
                                                                       not sexist
                                                                                            none
                                                                                                            none train
                         13149
                                    and over again, only with emojis
                                             instead of Morse code.
           sexism2022_english-
                                                           woman?
                                                                       not sexist
                                                                                            none
                                                                                                                   train
                                                                                                            none
                         13021
          sexism2022_english-
                                     I bet she wished she had a gun
                                                                       not sexist
                                                                                            none
                                                                                                            none
                                                                                                                    dev
                            966
          sexism2022_english-
                                 girls really get fucked almost every
                                                                       not sexist
                                                                                            none
                                                                                                            none train
                          5228
                                                        weekend?
                                        The hatred for moslems are
                                     Reasonable and Justified. They
                                     openly call for our subjugation,
          sexism2022_english-
                                       the rapes of our women and
                                                                       not sexist
                                                                                            none
                                                                                                            none train
                         10140
                                       children, our slavery, and our
                                     death. I'm not going to coexist
                                                        with them.
          sexism2022_english-
                                   Now this is a woman who gets it.
                                                                       not sexist
                                                                                            none
                                                                                                            none train
                          9726
                                  "American Idol" finalist [USER] said
                                   nothing is going to stop her from
          sexism2022_english-
                                       performing at The [USER] on
                                                                       not sexist
                                                                                            none
                                                                                                            none train
                         13365
                                     #IndependenceDay, calling the
                                        chance a once-in-a-lifetime
                                           opportunity. [URL] [URL]
                                                                                                    2.2 aggressive
          sexism2022_english-
                                   this bi--th should be stoped she's
```

sexist

the rapist

2. derogation

and emotive

attacks

test

20000 rows × 5 columns

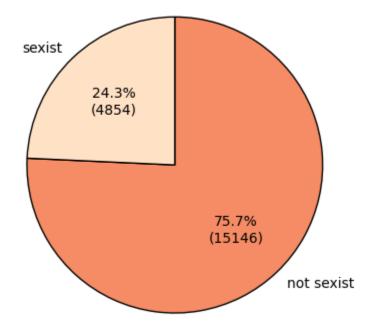
Labels Analysis

Setting up plot formatting

3523

Task A's labels distribution and counts

Task A labels distribution



5 random samples from the each label:

```
In []: # sexist samples:
    sexist = edos[edos['label_sexist'] == 'sexist'].sample(5).reset_index(drop=True)
    print('5 random sexist samples:')
    print(sexist['text'].values, '\n')

# not-sexist samples:
    not_sexist = edos[edos['label_sexist'] == 'not sexist'].sample(5).reset_index(drop=True)
    print('5 random not sexist samples:')
    print(not_sexist['text'].values)
```

5 random sexist samples:

['Ms...Fernandez killed her 8 year old son because she thought he was "gay" No words about a fathe r being involved with the child. How about having something other than a female around to mentor a nd direct a boy into manhood?'

'AWALT. The guy had social status, so she attached to it instantly.'

"Don't be sad - she is just a stupid 'ho."

'When it pertains to women, the size of her tits is directly proportional to the length of the hu g I give.'

'yeah it sucks being a white man now. We see inferiors promoted above us our women going off with savages, half breed children in tow. What do we have left to lose?']

5 random not sexist samples:

["they're all at chad's kegger tonight. they'll get back to LARPing tomorrow morning"

'Better way to repay the debt to us Americans, kick those mother fucking Muslims out of England. 100% of them need to be deported. Islam is the new Nazi.'

'Dude, i highly doubt i was better looking than you were at 15-20. i was a small acne ridden phag got and still made out with girls. Whats you excuse ?'

"This is what this sub should be about. Not complaining about your struggles or bragging about the girls you've been with."

'A woman wiling to pay her share is a woman expecting a much larger payoff in the near future.']

Task B's labels distribution and counts

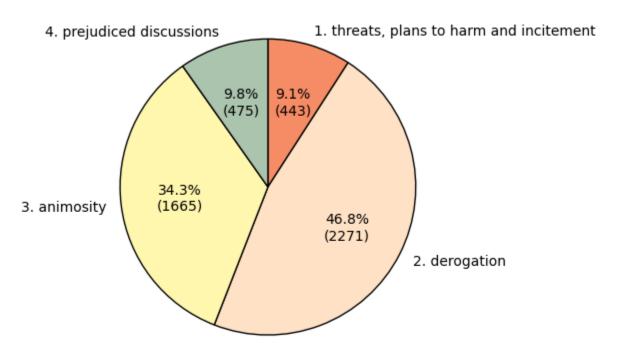
```
In [ ]: edos_sexist = edos[edos['label_sexist'] == 'sexist']
edos_sexist
```

\cap		+	Г	7
U	и	L	П	

rewire_id					-
sexism2022_english- 7228	[USER] Leg day is easy. Hot girls who wear miniskirts get asked out.	sexist	3. animosity	3.3 backhanded gendered compliments	train
sexism2022_english- 872	I get a new pussy every other week or whenever i want really. Didnt have to learn how to either from YouTube videos. Also PUAs dont get laid. At best they get 5/10 asian ladies after spending years pretending to be something they aren't. Its pathetic	sexist	2. derogation	2.3 dehumanising attacks & overt sexual objectification	train
sexism2022_english- 3553	I agree with that but at the same time I know myself well enough to say I can't love a woman. The minute she begins to hit the wall and some hotter, younger women enters the picture, it's time for impulse control because I'm going to want that.	sexist	2. derogation	2.1 descriptive attacks	dev
sexism2022_english- 14992	O come on there's no way any men are attracted to her shes a pig 🙄	sexist	2. derogation	2.3 dehumanising attacks & overt sexual objectification	train
sexism2022_english- 13328	democrats from Minnesota (leftist muzzie shithole) are the dumbest anti-americans there are. KLobuchar, one of the reps. is a dumb bitch who has a moronic position on the tax bill. Fuck this stupid woman	sexist	2. derogation	2.2 aggressive and emotive attacks	dev
•••					
sexism2022_english- 17659	This is easily the dumbest thing ever written. Hot pussy craves J. Todd Kincannon. It always has. It always will. I don't care if gross pussy doesn't like me. What do you think I am, some kind of psycho?	sexist	2. derogation	2.3 dehumanising attacks & overt sexual objectification	train
sexism2022_english- 13737	Recently some thots started advertizing their onlyfans on linkedin, i saw some posts of chat messages.	sexist	3. animosity	3.1 casual use of gendered slurs, profanities, and insults	train
sexism2022_english- 8065	I wish yall understand what a burden it can be for a woman to get pregnant or to have sex. So keep your legs closed, you dirty slut. Nobody is forcing you to fuck unemployed men who don't commit - that is 100% your choice.	sexist	2. derogation	2.2 aggressive and emotive attacks	train
sexism2022_english- 12176	idiocracy always seems to be female. Best evidence just this chick.	sexist	2. derogation	2.1 descriptive attacks	train
sexism2022_english- 3523	this bith should be stoped she's the rapist	sexist	2. derogation	2.2 aggressive and emotive attacks	test

text label_sexist label_category label_vector split

Task B labels distribution



Sample for each label:

```
In [ ]: for label in b_labels:
    print('Label:', label)
    print(edos_sexist[edos_sexist['label_category'] == label]['text'].sample(3).values, '\n')
```

Label: 1. threats, plans to harm and incitement

['Just convert to Islam and shout Allahu Akbar before you assault woman from now on. If your a muz zie they will literally give you a medal and they keys to the city when you throw acid in a whores face.'

"If it was in where I live. You'd give her series of slaps to the cheek and still not get arreste d because she hit you first. You had the right to self defence man!"

'Next time a castrationista goes off on another men are dirt, men are trash, men are scum, men ar e pigs rant, stuff that in her face.']

Label: 2. derogation

['Why is she in cadet uniform? TAKE HER ASS BACK TO THE KITCHEN!'

'And every time I see a lib tart bitch with pink hair I wish INSANE ASYLUMS TO COME BACK!'

'This true . Obese hideous women and stick thin men without an ounce of muscle or fat.']

Label: 3. animosity

["lol it's cool to bitch every now and then I'm a dude and I sometimes bitch more than a woman but I know where I get it's from my Dad he bitches more than any human can possibly to bitch he will d ie bitching and also be bitching in death"

'Those guys couldnt care less for average and below average women. Why would they even bother with ugly women when they can get 10/10 virgin models?'

'ed sheeran no wonder they stole your nhs records it says born a she you are a bitch your hips do nt lie UK is full of trans']

```
Label: 4. prejudiced discussions
```

["women's suffrage was a mistake" 'First and foremost, fire all women.'

'Well, I learned something new today. Did she divorce rape him (resistance is futile)']

```
In [ ]: c_labels, c_counts = np.unique(edos_sexist['label_vector'], return_counts=True)
         c_labels, c_counts = sorted(c_labels), c_counts[np.argsort(c_labels)]
         c_{sum} = sum(c_{counts})
         fig, ax = plt.subplots()
         bars = ax.barh(width=c_counts,
                         y=c_labels,
                         color=colors,
                         edgecolor='black')
         # format labels: value, percentage = value/sum
         def bar_format(values):
             def my_format(value):
                 val = int(round(value))
                 \label{lem:condition} \textbf{return '}\{v:d\} \setminus n(\{p:.1f\}\%)'. format(v=val, p=val/c\_sum*100)
             return my_format
         ax.bar_label(bars, fmt=bar_format(c_counts),
                       label_type='edge',
                       padding=5)
         ax.margins(x=0.15)
         ax.invert_yaxis()
         plt.title("Task C labels distribution", fontdict=dict(weight='bold', size=14))
         plt.savefig('task_c_labels_distribution.png', dpi=300, bbox_inches='tight')
         plt.show()
```

Task C labels distribution 1.1 threats of harm (1.6%)363 1.2 incitement and encouragement of harm (7.5%)1024 2.1 descriptive attacks (21.1%)961 2.2 aggressive and emotive attacks (19.8%)286 2.3 dehumanising attacks & overt sexual objectification (5.9%)910 3.1 casual use of gendered slurs, profanities, and insults (18.7%) 596 3.2 immutable gender differences and gender stereotypes (12.3%)91 3.3 backhanded gendered compliments (1.9%) 68 3.4 condescending explanations or unwelcome advice (1.4%) 4.1 supporting mistreatment of individual women (2.2%)368 4.2 supporting systemic discrimination against women as a group (7.6%)

200

400

600

800

1000

Sample for each label:

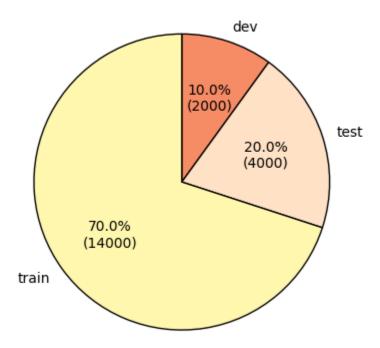
```
In [ ]: for label in c_labels:
    print('Label:', label)
    print(edos_sexist[edos_sexist['label_vector'] == label]['text'].sample(1).values, '\n')
```

```
Label: 1.1 threats of harm
["As a kid, I grew up in my teen years on the streets. She wouldn't have succeed in hitting more t
han twice before I beat the living hell out of her."]
Label: 1.2 incitement and encouragement of harm
['There should be a cost to those who purposely lie about other politicians in order to defeat the
m. Old hag Feinstein should fry']
Label: 2.1 descriptive attacks
["This kind of protest makes sense to women because they spend their entire lives using their bodi
es to get whatever they want. It's all they know how to do. Logical arguments? Nah. I'll just take
my clothes off and shout."]
Label: 2.2 aggressive and emotive attacks
['she is a 24 carat cunt.']
Label: 2.3 dehumanising attacks & overt sexual objectification
["It's not his job to keep other guys away, it's her fucking job, and if she doesn't respect that
he gotta get the fuck out. A billion pussy in the sea"]
Label: 3.1 casual use of gendered slurs, profanities, and insults
['She's either an attention whore or she had a better option come along. Don't sweat it, there's a
lways more.']
Label: 3.2 immutable gender differences and gender stereotypes
["Well I live at my gym, and women are the same everywhere, so it's almost as if I do live at your
gym."]
Label: 3.3 backhanded gendered compliments
['Thank you for all the women who are still sensible.']
Label: 3.4 condescending explanations or unwelcome advice
['little girl, youre really the boss--come back tomorrow and find out how']
Label: 4.1 supporting mistreatment of individual women
['Odin raped all those giant women and mortal women with their consent.']
Label: 4.2 supporting systemic discrimination against women as a group
['Several women out there that got a lesser sentence for murder.']
```

Data Splitting Distribution

Train, Dev, Test split distribution:

Dataset splits distribution

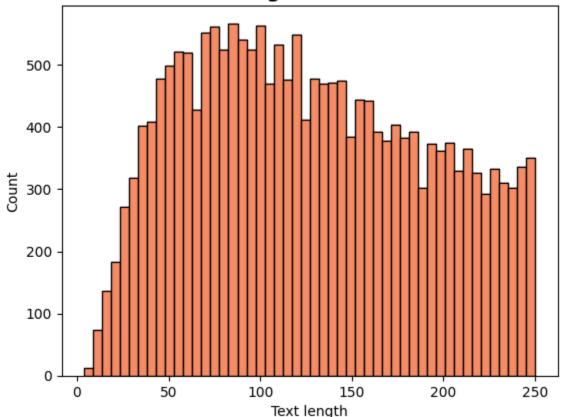


Sentence Length Analysis

Max, min, mean, median, and standard deviation of sentence length of all data

```
In [ ]: edos['text_len'] = edos['text'].apply(lambda x: len(x))
        edos['text_len'].astype(int)
        edos['text_len'].describe()
Out[]: count
                 20000.000000
        mean
                   126.996200
        std
                    63.256474
        min
                    4.000000
                    74.000000
        25%
        50%
                   121.000000
                   178.250000
                   250.000000
        Name: text_len, dtype: float64
        Sentence length histogram:
In [ ]: plt.hist(edos['text_len'], bins=50,
                 edgecolor='black')
        plt.title("Text length distribution", fontdict=dict(weight='bold', size=14))
        plt.xlabel('Text length')
        plt.ylabel('Count')
        plt.savefig('text_length_distribution.png', dpi=300, bbox_inches='tight')
        plt.show()
```

Text length distribution



Example of the longest sentence and the shortest sentence:

Number of samples with text length equal to max: 63

```
In [ ]: print('Shortest text:')
    print(edos[edos['text_len'] == edos['text_len'].min()]['text'].sample(1).values, '\n')
    print('Longest text:')
    print(edos[edos['text_len'] == edos['text_len'].max()]['text'].sample(1).values)

Shortest text:
```

SHOPLEST TEXT.

['Zits']

Longest text:

['The assassination of Archduke Franz Ferdinand of Austria,(shown here) heir presumptive to the Austro-Hungarian throne, and his wife Sophie, Duchess of Hohenberg, occurred on 28 June 1914 in Sara jevo. It was used as the pretext to start World War One.']

Sentence length by each split description

```
In [ ]: edos.groupby('split')['text_len'].describe()
Out[ ]:
                                                          50%
                 count
                                          std min 25%
                                                                 75%
                             mean
                                                                       max
         split
                2000.0
                       125.430000
                                   63.477349
                                                    74.0
                                                         119.0
                                                               176.0
                                                                      250.0
          dev
          test
                4000.0
                       125.442750
                                   63.820910
                                               5.0
                                                    72.0
                                                         120.0
                                                               177.0 250.0
               14000.0 127.663786 63.055376
                                                   75.0 122.0 179.0 250.0
                                               4.0
         train
```

Sentence length by each task A label

```
In [ ]: edos.groupby('label_sexist')['text_len'].describe()
```

```
        Out[]:
        count
        mean
        std
        min
        25%
        50%
        75%
        max

        label_sexist

        not sexist
        15146.0
        124.394890
        63.302372
        4.0
        71.0
        118.0
        175.0
        250.0

        sexist
        4854.0
        135.113103
        62.426448
        9.0
        83.0
        132.0
        188.0
        250.0
```

Extract 10 shortest sentences for 'not sexist' label:

```
In [ ]: edos[edos['label_sexist'] == 'not sexist'].sort_values(by='text_len').head(10).to_csv('10 shorte:
```

Sentence length by each task B label

```
In []: edos_sexist['text_len'] = edos_sexist['text'].apply(lambda x: len(x))
    edos_sexist

C:\Users\Minh Ha\AppData\Local\Temp\ipykernel_18368\2283281652.py: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-a-copy
    edos_sexist['text_len'] = edos_sexist['text'].apply(lambda x: len(x))
```

Out[]: text label_sexist label_category label_vector split text_len

rewire_id						
sexism2022_english- 7228	[USER] Leg day is easy. Hot girls who wear miniskirts get asked out.	sexist	3. animosity	3.3 backhanded gendered compliments	train	68
sexism2022_english- 872	I get a new pussy every other week or whenever i want really. Didnt have to learn how to either from YouTube videos. Also PUAs dont get laid. At best they get 5/10 asian ladies after spending years pretending to be something they aren't. Its pathetic	sexist	2. derogation	2.3 dehumanising attacks & overt sexual objectification	train	250
sexism2022_english- 3553	I agree with that but at the same time I know myself well enough to say I can't love a woman. The minute she begins to hit the wall and some hotter, younger women enters the picture, it's time for impulse control because I'm going to want that.	sexist	2. derogation	2.1 descriptive attacks	dev	244
sexism2022_english- 14992	O come on there's no way any men are attracted to her shes a pig 🙄	sexist	2. derogation	2.3 dehumanising attacks & overt sexual objectification	train	65
sexism2022_english- 13328	democrats from Minnesota (leftist muzzie shithole) are the dumbest anti- americans there are. KLobuchar, one of the reps. is a dumb bitch who has a moronic position on the tax bill. Fuck this stupid woman	sexist	2. derogation	2.2 aggressive and emotive attacks	dev	203
sexism2022_english- 17659	This is easily the dumbest thing ever written. Hot pussy craves J. Todd Kincannon. It always has. It always will. I don't care if gross pussy doesn't like me. What do you think I am, some kind of psycho?	 sexist	 2. derogation	2.3 dehumanising attacks & overt sexual objectification	 train	203
sexism2022_english- 13737	Recently some thots started advertizing their onlyfans on linkedin, i saw some posts of chat messages.	sexist	3. animosity	3.1 casual use of gendered slurs, profanities, and insults	train	102
sexism2022_english- 8065	I wish yall understand what a burden it can	sexist	2. derogation	2.2 aggressive and emotive	train	222

be for a woman to get

attacks

text	label_sexist	label_category	label_vector	split	text_len
------	--------------	----------------	--------------	-------	----------

•	
rewire	ıd

pregnant or to have sex. So keep your legs closed, you dirty slut. Nobody is forcing you to fuck unemployed men who don't commit - that is 100%your choice. idiocracy always sexism2022_englishseems to be female. 2.1 descriptive 67 sexist 2. derogation train 12176 Best evidence just this attacks chick. 2.2 aggressive sexism2022_englishthis bi--th should be sexist 2. derogation and emotive 45 test 3523 stoped she's the rapist attacks

4854 rows × 6 columns

In []: edos_sexist.groupby('label_category')['text_len'].describe()

Dut[]:		count	mean	std	min	25%	50%	75%	max
	label_category								
	1. threats, plans to harm and incitement	443.0	120.234763	63.519114	9.0	66.0	110.0	173.0	250.0
	2. derogation	2271.0	135.432849	61.410181	12.0	85.0	133.0	187.0	250.0
	3. animosity	1665.0	132.958559	63.119850	10.0	80.0	130.0	185.0	250.0
	4. prejudiced discussions	475.0	155.012632	58.906597	14.0	113.0	157.0	205.5	250.0

Sentence length by each task C label

	5 ,								
[]: e	dos_sexist.groupby('label_vector')['text	_len'].desc	ribe()					
]:		count	mean	std	min	25%	50%	75%	max
	label_vector								
	1.1 threats of harm	80.0	125.037500	55.474123	32.0	80.75	118.5	168.50	249.0
	1.2 incitement and encouragement of harm	363.0	119.176309	65.180640	9.0	64.50	108.0	176.50	250.0
	2.1 descriptive attacks	1024.0	144.484375	61.121770	21.0	93.75	143.0	197.00	249.0
	2.2 aggressive and emotive attacks	961.0	127.940687	60.686744	12.0	78.00	123.0	174.00	250.0
	2.3 dehumanising attacks & overt sexual objectification	286.0	128.199301	60.711617	27.0	78.00	119.5	178.75	250.0
	3.1 casual use of gendered slurs, profanities, and insults	910.0	127.373626	65.088963	10.0	72.00	122.5	182.00	250.0
3	3.2 immutable gender differences and gender stereotypes	596.0	140.612416	59.675333	20.0	92.00	135.0	191.00	250.0
	3.3 backhanded gendered compliments	91.0	128.615385	59.243522	22.0	86.00	125.0	177.00	248.0
	3.4 condescending explanations or unwelcome advice	68.0	146.426471	63.057736	29.0	95.75	152.0	194.00	250.0
	4.1 supporting mistreatment of individual women	107.0	164.046729	56.526231	50.0	120.00	161.0	208.50	249.0
	4.2 supporting systemic discrimination against women as a	368.0	152.385870	59.397731	14.0	108.75	155.5	205.00	250.0

group

out[]:			text	label_sexist	label_category	split
	label_vector	rewire_id				
	1.1 threats of harm	sexism2022_english- 7329	l'd love to fucking kick her ass	sexist	1. threats, plans to harm and incitement	train
		sexism2022_english- 18782	No poppy? You deserve an acid attack.	sexist	1. threats, plans to harm and incitement	train
		sexism2022_english- 13254	This woman needs a rope around her neck	sexist	1. threats, plans to harm and incitement	train
		sexism2022_english- 3313	I would kill your mother to get a shot at you.	sexist	1. threats, plans to harm and incitement	train
		sexism2022_english- 9066	Stupid bitch I would've slapped tf out of her	sexist	1. threats, plans to harm and incitement	train
						•••
	4.2 supporting systemic discrimination against women as a group	sexism2022_english- 2034	First and foremost, fire all women.	sexist	4. prejudiced discussions	train
		sexism2022_english- 16219	Don't pay attention to what women say.	sexist	4. prejudiced discussions	test
		sexism2022_english- 8735	She got her equal rights And lefts 😂	sexist	4. prejudiced discussions	train
		sexism2022_english- 16219 what women say. She got her equal rights And lefts Sexism2022_english- 19085 This is why women should have no rights.	4. prejudiced discussions	dev		
		sexism2022_english- 10350	To female Bishops stop making fun of religion.	sexist	4. prejudiced discussions	test

110 rows × 4 columns

Word Count Analysis

Word count of all data description

```
In []: import preprocess

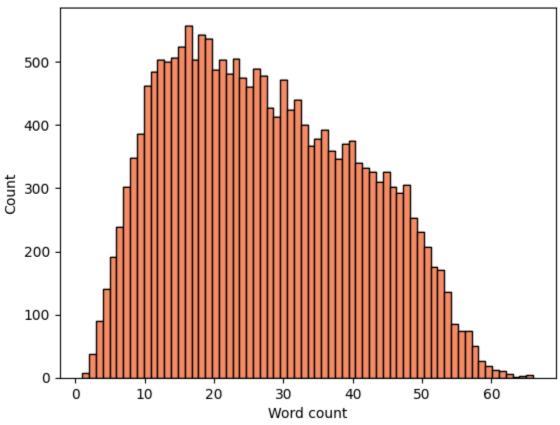
Import English dictionary

In []: english_words = set()
    with open('../wiki_dict.txt', 'r', encoding='utf-8') as f:
        for line in f:
            english_words.add(line.strip())
    len(english_words)
```

```
return preprocess.tokenize(text, vocab_set=english_words)
          print(edos['text'][539])
          print(tokenize(edos['text'][539]))
        ▶Anonymous 06/20/18 (Wed) 14:05:56 0f806a (15) No.1831793 =NEWS= World Cup reporter sexually assa
        ulted during live broadcast www.foxnews.com/entertainment/2018/06/20/female-world-cup-reporter-sex
        ually-assaulted-during-live-broadcast.html
        ['\blacktriangleright', 'anonymous', '06/20', '/', '18', '(', 'wed', ')', '14:05', '56', '0f806a', '(', '15', ')', 'no', '.', '1831793', 'news', 'world', 'cup', 'reporter', 'sexually', 'assaulted', 'during', 'liv
        e', 'broadcast', 'url']
        ['▶', 'anonymous', '06/20', '/', '18', '(', 'wed', ')', '14:05', '56', '0f806a', '(', '15', ')', 'no', '.', '1831793', 'news', 'world', 'cup', 'reporter', 'sexually', 'assaulted', 'during', 'liv
        e', 'broadcast', 'url']
In [ ]: edos['word_count'] = edos['text'].apply(lambda x: len(tokenize(x)))
          edos['word_count'].astype(int)
          edos['word_count'].describe()
Out[]: count
                     20000.000000
          mean
                        27,296200
                        13,666726
          std
          min
                         1.000000
          25%
                        16.000000
          50%
                        26.000000
          75%
                        38.000000
                        66,000000
          max
          Name: word_count, dtype: float64
          Histogram of word count:
In [ ]: plt.hist(edos['word_count'], bins=edos['word_count'].max(),
                         edgecolor='black')
          plt.title("Word count distribution", fontdict=dict(weight='bold', size=14))
          plt.xlabel('Word count')
          plt.ylabel('Count')
          plt.savefig('word_count_distribution.png')
          plt.show()
```

In []: def tokenize(text):

Word count distribution



Word count by each split description

```
In [ ]: edos.groupby('split')['word_count'].describe()
```

```
split
                2000.0 26.977500 13.741885
                                             1.0
                                                 16.0
                                                      25.0
                                                            37.0
                                                                  63.0
          dev
                4000.0 27.009500 13.789868
                                                 16.0
                                                       25.0
                                                            38.0
                                                                  66.0
                                             1.0
          test
              14000.0 27.423643 13.619539
                                            1.0
                                                 16.0
                                                      26.0
                                                            38.0
                                                                  66.0
         train
         Word count by each task A label
In [ ]: edos.groupby('label_sexist')['word_count'].describe()
Out[ ]:
                                             std min 25% 50% 75% max
                      count
                                mean
         label_sexist
          not sexist 15146.0 26.743563 13.672744
                                                   1.0
                                                       15.0
                                                             25.0
                                                                  37.0
                                                                        66.0
              sexist
                     4854.0 29.020602 13.504706
                                                  2.0
                                                       18.0
                                                             28.0
                                                                  40.0
                                                                        66.0
In [ ]: # shortest and longest texts for each label
         for label in edos['label_sexist'].unique():
             print('Label:', label)
             print('Shortest text:', edos[edos['label_sexist'] == label].sort_values(by='word_count').head
             print('Longest text:', edos[edos['label_sexist'] == label].sort_values(by='word_count', ascer
       Label: not sexist
       Shortest text: ['wabbits']
       Longest text: ['"Nit(s) is a common name of a head lice (louse) egg. A female louse can lay up to
       10 nits per day, and it usually takes about 7-10 days for them to hatch. If left untreated in a 30
       -day lifespan a louse can lay up to 100 eggs in her lifetime." [URL]']
       Label: sexist
       Shortest text: ['Hoochie mama']
       Longest text: ["Gjdm, I need to stop being a bitch and talk to woman I like during the day. Heck,
       there's a girl at my gym who I've been dying to strike up a conversation with but I always make an
       excuse in my head not to. F that . Even it's just 'hi' I'm doing it."]
         Word count by each task B label
In [ ]: edos_sexist['word_count'] = edos_sexist['text'].apply(lambda x: len(tokenize(x)))
       C:\Users\Minh Ha\AppData\Local\Temp\ipykernel_18368\3223232917.py: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/inde
       xing.html#returning-a-view-versus-a-copy
         edos_sexist['word_count'] = edos_sexist['text'].apply(lambda x: len(tokenize(x)))
In [ ]: | edos_sexist.groupby('label_category')['word_count'].describe()
Out[ ]:
                                                                     std min 25% 50% 75% max
                                              count
                                                         mean
                               label_category
                                               443.0 26.510158 13.848951
                                                                               15.0
                                                                                     25.0
                                                                                           38.0
                                                                                                66.0
         1. threats, plans to harm and incitement
                                                                           3.0
                                2. derogation 2271.0 29.021136 13.259540
                                                                           2.0
                                                                               18.0
                                                                                     28.0
                                                                                           39.0
                                                                                                65.0
                                 3. animosity 1665.0 28.763363 13.841930
                                                                           2.0
                                                                               17.0
                                                                                     27.0
                                                                                           40.0
                                                                                                66.0
                      4. prejudiced discussions
                                              475.0 32.261053 12.536161
                                                                           3.0
                                                                               23.0
                                                                                     32.0
                                                                                          42.0
                                                                                                62.0
In [ ]: # shortest and longest texts for each label
         for label in sorted(edos_sexist['label_category'].unique()):
             print('Label:', label)
             print('Shortest text:', edos_sexist[edos_sexist['label_category'] == label].sort_values(by=')
```

print('Longest text:', edos_sexist[edos_sexist['label_category'] == label].sort_values(by='wo')

25% 50% 75%

Out[]:

count

Label: 1. threats, plans to harm and incitement

Shortest text: ['Hang her.']

Longest text: ["I would just start fucking her!! What's your problem, bro? You don't ask, you just do it!! In my 42 years on this earth I've never had a female that wasn't a relative sleep over at my crib/pad and not give me some pussy!! You are simping big time!!"]

Label: 2. derogation

Shortest text: ['Hoochie mama']

Longest text: ["ITT: 1. She's not that hot. I would not bang her. I know hotter women. 2. She's a bitch. I would not bang her, out of principle. Because I only bang girls with *inner beauty*. 3. Of course, she get's hit on. She's just saying that she doesn't."]

Label: 3. animosity

Shortest text: ['Worthless bitch']

Longest text: ["Gjdm, I need to stop being a bitch and talk to woman I like during the day. Heck, there's a girl at my gym who I've been dying to strike up a conversation with but I always make an excuse in my head not to. F that . Even it's just 'hi' I'm doing it."]

Label: 4. prejudiced discussions Shortest text: ['We ban females']

Longest text: ['All depends on the divorce agreement, or if you go to court, depends on the state you live in, and the judge you get. And by "depends," I mean "how bad it will be for YOU, the man. It WILL be bad, but just how bad depends on several factors.']

Word count by each task C label

```
In [ ]: edos_sexist.groupby('label_vector')['word_count'].describe()
Out[]:
                                                                                          25%
                                                                                                50%
                                                                                                       75%
                                                     count
                                                                                   min
                                       label_vector
                                                       80.0 28.762500 12.471988
                                                                                        19.00
                                1.1 threats of harm
                                                                                    8.0
                                                                                                27.0 38.00
                                                                                                             57.0
               1.2 incitement and encouragement of
                                                      363.0 26.013774 14.101931
                                                                                    3.0
                                                                                         14.00
                                                                                                24.0 37.00
                                                                                                             66.0
                                              harm
                             2.1 descriptive attacks
                                                     1024.0 30.511719 13.210450
                                                                                    4.0
                                                                                         20.00
                                                                                                30.0 41.00
                                                                                                             63.0
                                                                                         17.00
                 2.2 aggressive and emotive attacks
                                                      961.0 27.718002
                                                                        13.200481
                                                                                    2.0
                                                                                                27.0
                                                                                                     38.00
                                                                                                             65.0
            2.3 dehumanising attacks & overt sexual
                                                      286.0 28.062937
                                                                        13.128096
                                                                                    5.0 17.25
                                                                                                26.0 38.00
                                                                                                             58.0
                                     objectification
                    3.1 casual use of gendered slurs,
                                                      910.0 27.582418
                                                                        14.171095
                                                                                         15.25
                                                                                                27.0
                                                                                                     40.00
                                                                                    2.0
                                                                                                             66.0
                             profanities, and insults
              3.2 immutable gender differences and
                                                      596.0 30.380872 13.124135
                                                                                         20.00
                                                                                                29.5
                                                                                                     40.00
                                                                                                             60.0
                                gender stereotypes
             3.3 backhanded gendered compliments
                                                       91.0 28.241758
                                                                        13.514881
                                                                                         18.00
                                                                                                      38.00
                                                                                    5.0
                                                                                                26.0
                                                                                                             57.0
                 3.4 condescending explanations or
                                                       68.0 31.088235
                                                                        14.522998
                                                                                    5.0
                                                                                         18.75
                                                                                                29.5
                                                                                                     42.00
                                                                                                             66.0
                                 unwelcome advice
           4.1 supporting mistreatment of individual
                                                      107.0 34.691589 12.378779
                                                                                   11.0 26.00
                                                                                                34.0 44.50
                                                                                                             62.0
                                           women
             4.2 supporting systemic discrimination
                                                      368.0 31.554348 12.509714
                                                                                    3.0 22.00
                                                                                                32.0 41.25
                         against women as a group
```

```
In []: # shortest and Longest texts for each Label
for label in sorted(edos_sexist['label_vector'].unique()):
    print('Label:', label)
    print('Shortest text:', edos_sexist[edos_sexist['label_vector'] == label].sort_values(by='work
    print('Longest text:', edos_sexist[edos_sexist['label_vector'] == label].sort_values(by='work
```

Label: 1.1 threats of harm

Shortest text: ["I'd love to fucking kick her ass"]

Longest text: ['Got no response to that. Oh great the French view is what exactly? hes a black guy too. Holy shit there is nothing about this post that doesnt want me to punch her fat American fac e. Yes i assume shes fat, thats what you get from stupid assumptions.']

Label: 1.2 incitement and encouragement of harm

Shortest text: ['Hang her.']

Longest text: ["I would just start fucking her!! What's your problem, bro? You don't ask, you just do it!! In my 42 years on this earth I've never had a female that wasn't a relative sleep over at my crib/pad and not give me some pussy!! You are simping big time!!"]

Label: 2.1 descriptive attacks

Shortest text: ['Israeli women fuck donkeys']

Longest text: ['1. They're looking to see if other women want you (social proof) so they don't see m weird if they want you. 2. If you're a virgin, lie. Fake it till you make it. The girl who I los t my virginity to thought I had slept with 7 women (LOL).']

Label: 2.2 aggressive and emotive attacks

Shortest text: ['Hoochie mama']

Longest text: ["ITT: 1. She's not that hot. I would not bang her. I know hotter women. 2. She's a bitch. I would not bang her, out of principle. Because I only bang girls with *inner beauty*. 3. O f course, she get's hit on. She's just saying that she doesn't."]

Label: 2.3 dehumanising attacks & overt sexual objectification

Shortest text: ['Chad is every foids soulmate']

Longest text: ['Are these people retarded? 16 y/o girls are cock-ready. It can be 30 y/o cock of a respectable, responsible prosecutor who wants to marry them or it can be 18 y/o cock attached to a n irresponsible guy who may be a mechanic or something one day. [URL]']

Label: 3.1 casual use of gendered slurs, profanities, and insults

Shortest text: ['Worthless bitch']

Longest text: ["Gjdm, I need to stop being a bitch and talk to woman I like during the day. Heck, there's a girl at my gym who I've been dying to strike up a conversation with but I always make an excuse in my head not to. F that . Even it's just 'hi' I'm doing it."]

Label: 3.2 immutable gender differences and gender stereotypes

Shortest text: ['typical female nature']

Longest text: ['I\'m a 5\'6 latino with a 5" dick. Would I be able to get a low-end (1-3/10) white woman in the US if I careermaxx? Also, is it easier for latinos to integrate and be accepted by wh ite locals than other races like blacks or asians?']

Label: 3.3 backhanded gendered compliments

Shortest text: ['Strong women scare me.']

Longest text: ["Exactly. 70's bush was just wild. No one is saying it has to be shaved... but extremely hairy and unmanaged isn't a great thing to most men. If it were true, that would mean basic hygiene wouldn't be an issue and we would be cool with ugly women."]

Label: 3.4 condescending explanations or unwelcome advice

Shortest text: ['Ladies.... never get tattoos.']

Longest text: ['Well if you have a penis you have to wear skinny jeans and be a beta, if you own a vagina, don\'t wash it for a week and bow up and act tough BUT be sure to scream "you hit a girl" if you\'re punched. Oh and if you see a car coming MOVE OUT OF THE WAY!']

Label: 4.1 supporting mistreatment of individual women

Shortest text: ['She disobeyed her male keeper. She knew the rules.']

Longest text: ["Sell it to who pays more man. chicks come and go cash is forever. plus it'll tell her you arent a simp who will do anything for the pussy. side note, i dont know how you put up wit h chicks that talk in slang or like a dude, bugs the shit out of me"]

Label: 4.2 supporting systemic discrimination against women as a group

Shortest text: ['We ban females']

Longest text: ['All depends on the divorce agreement, or if you go to court, depends on the state you live in, and the judge you get. And by "depends," I mean "how bad it will be for YOU, the man. It WILL be bad, but just how bad depends on several factors.']

Word Analysis

Get word frequency of all data

```
In [ ]: # get all texts id for every text that contains word
        def get_text_ids(word):
            return edos[edos['tokenized_text'].apply(lambda x: word in x)].index.values
In [ ]: words = pd.DataFrame(columns=['word'])
        edos['tokenized_text'] = edos['text'].apply(lambda x: tokenize(x))
        words['word'] = edos['tokenized_text'].explode().unique()
        words
Out[]:
                   word
             0
                     in
             1
                 nigeria
             2
             3
                      if
             4
                    you
         25316 selflessly
         25317
         25318
                    idol
         25319
                  finalist
         25320
                 stoped
        25321 rows × 1 columns
In [ ]: words.sort_values(by='word', ascending=True).reset_index(drop=True)
Out[]:
                word
             0
                    ļ
             1
             2
                   #
             4
                #cw2
         25316
         25317
         25318
         25319
         25320
        25321 rows × 1 columns
        Get word frequency
In [ ]: from nltk import FreqDist
        word_freq = FreqDist(edos['tokenized_text'].explode())
        words['freq'] = words['word'].apply(lambda x: word_freq.get(x))
In [ ]: # most frequent words including stopwords
        words.sort_values(by='freq', ascending=False).reset_index(drop=True).head(20)
```

```
word
             freq
0
            27033
1
       the 14220
2
         , 14208
3
         a 12455
4
        to 12197
5
           11284
6
             9557
7
       and
             9194
8
       you
             8490
9
       not
             7507
10
             6732
        of
11
       she
             6161
12
      that
             6152
13
             5966
       her
14
         it
             5690
15
       are
             5293
16
        in
             5004
17 women
             4334
18
             4257
19
             4237
```

Out[]:

```
In []: # most frequent words excluding stopwords
    from nltk.corpus import stopwords
    from string import punctuation

    stop_words_punct = set(stopwords.words('english'))
    stop_words_punct.update(punctuation)

# remove stopwords and punctuation
    words['word'] = words['word'].apply(lambda x: x if (x not in stop_words_punct) and (not str(x).iccords_dropna(inplace=True))
    words.dropna(inplace=True)
    words['freq'] = words['word'].apply(lambda x: word_freq.get(x))

words.sort_values(by='freq', ascending=False).reset_index(drop=True).head(20)
```

```
1
                 url 3414
          2
                like 2429
              would 2063
          3
          4
                get 1707
          5
            woman 1643
          6
               men
                    1464
          7
               user 1363
          8
               one 1145
          9
               girls 1135
         10
                girl 1078
         11
               fuck 1073
         12
               want
                    1062
         13
                      979
              know
         14
              white
                      965
         15
                shit
                      957
         16
               think
                      954
         17
             people
                      953
         18
             female
                      951
         19
               even
                      909
In [ ]: words['freq'].describe()
Out[]: count
                  24902.000000
        mean
                     9.806401
                     60.593327
        std
                      1.000000
        min
        25%
                      1.000000
        50%
                      2.000000
        75%
                      4.000000
                   4334.000000
        max
        Name: freq, dtype: float64
In [ ]: plt.hist(words['freq'], bins=50)
        plt.title("Word frequency distribution", fontdict=dict(weight='bold', size=14))
```

plt.savefig('word_frequency_distribution.png', dpi=300, bbox_inches='tight')

Out[]:

word

0 women

freq

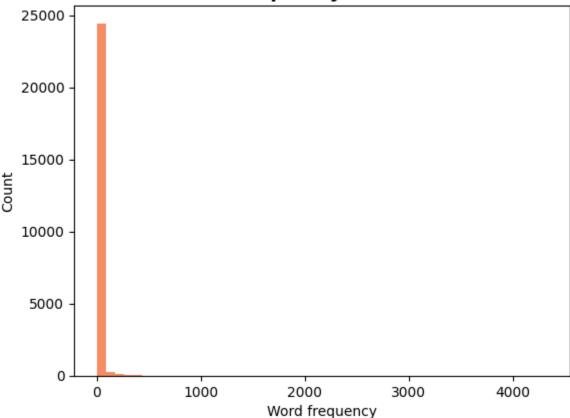
4334

plt.xlabel('Word frequency')

plt.ylabel('Count')

plt.show()

Word frequency distribution



```
print("Number of words with frequency = 1:", words[words['freq'] == 1].shape[0])
         print("Number of words with frequency < 10:", words[words['freq'] < 10].shape[0])
print("Number of words with frequency > 100:", words[words['freq'] > 100].shape[0])
         print("Number of words with frequency > 1000:", words[words['freq'] > 1000].shape[0])
       Number of words with frequency = 1: 12408
       Number of words with frequency < 10: 21462
       Number of words with frequency > 100: 388
       Number of words with frequency > 1000: 13
         Get word frequency of each label
In [ ]: edos_sexist['tokenized_text'] = edos_sexist['text'].apply(lambda x: tokenize(x))
         word_freq_sexist = FreqDist(edos_sexist['tokenized_text'].explode())
         words['freq_sexist'] = words['word'].apply(lambda x: word_freq_sexist.get(x)).fillna(0)
         words['freq_sexist'] = words['freq_sexist'].astype(int)
         words['freq_not_sexist'] = words['freq'] - words['freq_sexist']
       C:\Users\Minh Ha\AppData\Local\Temp\ipykernel_18368\1519064831.py: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/inde
       xing.html#returning-a-view-versus-a-copy
         edos_sexist['tokenized_text'] = edos_sexist['text'].apply(lambda x: tokenize(x))
```

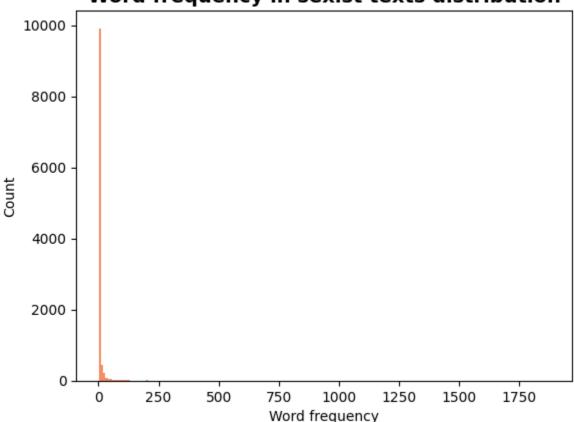
In []: words.sort_values(by='freq_sexist', ascending=False).reset_index(drop=True).head(20)

	word	freq	freq_sexist	freq_not_sexist
0	women	4334	1877	2457
1	like	2429	697	1732
2	men	1464	658	806
3	bitch	591	577	14
4	get	1707	537	1170
5	woman	1643	535	1108
6	would	2063	529	1534
7	url	3414	507	2907
8	female	951	326	625
9	fuck	1073	326	747
10	want	1062	306	756
11	fucking	860	291	569
12	man	876	289	587
13	girls	1135	272	863
14	shit	957	269	688
15	one	1145	267	878
16	know	979	265	714
17	even	909	244	665
18	pussy	250	224	26
19	white	965	221	744

Out[]:

```
In [ ]: words[words['freq_sexist'] > 0]['freq_sexist'].describe()
Out[]: count
                 10914.000000
        mean
                     5.781473
                    28.312850
        std
                     1.000000
        min
        25%
                     1.000000
        50%
                     1.000000
        75%
                     3.000000
                  1877.000000
        max
        Name: freq_sexist, dtype: float64
In [ ]: plt.hist(words[words['freq_sexist'] > 0]['freq_sexist'], bins=200)
        plt.title("Word frequency in sexist texts distribution", fontdict=dict(weight='bold', size=14))
        plt.xlabel('Word frequency')
        plt.ylabel('Count')
        plt.savefig('word_frequency_sexist_distribution.png', dpi=300, bbox_inches='tight')
        plt.show()
```

Word frequency in sexist texts distribution



```
In [ ]: print("Number of words in sexist text with frequency = 1:", words[words['freq_sexist'] == 1].shaprint("Number of words in sexist text with frequency < 10:", words[words['freq_sexist'] < 10].shaprint("Number of words in sexist text with frequency > 100:", words[words['freq_sexist'] > 100].sprint("Number of words in sexist text with frequency > 1000:", words[words['freq_sexist'] > 1000].sprint("Number of words in sexist text with frequency = 1: 5844
Number of words in sexist text with frequency < 10: 23817
Number of words in sexist text with frequency > 100: 76
Number of words in sexist text with frequency > 100: 1
```

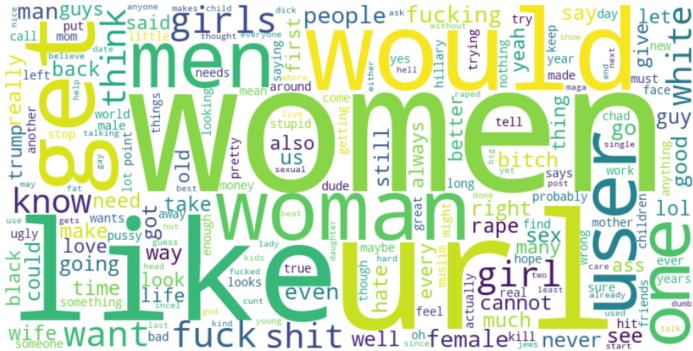
Create word cloud

```
In [ ]: from wordcloud import WordCloud
        def show_word_cloud(text: any, freq: any, title: str):
            text = np.array(text, dtype=str)
            freq = np.array(freq, dtype=int)
            wordcloud = WordCloud(width=800,
                                  height=400,
                                  background_color='white',
                                   collocations=False,
                                   relative scaling=0.5,
                                  max_words=200,
                                   ).generate_from_frequencies(dict(zip(text, freq)))
            plt.figure(figsize=(12, 6))
            plt.imshow(wordcloud, interpolation='bilinear')
            plt.axis('off')
            plt.title(title, fontdict=dict(weight='bold', size=14))
            plt.savefig(title.replace(' ', '_').lower() + '.png', dpi=300, bbox_inches='tight') # Lower
            plt.show()
```

Word cloud of all data

```
In [ ]: words = words.sort_values(by='freq', ascending=False)
    show_word_cloud(words['word'], words['freq'], 'Word cloud for all texts')
```

Word cloud for all texts



Word cloud of most frequent words of sexist label

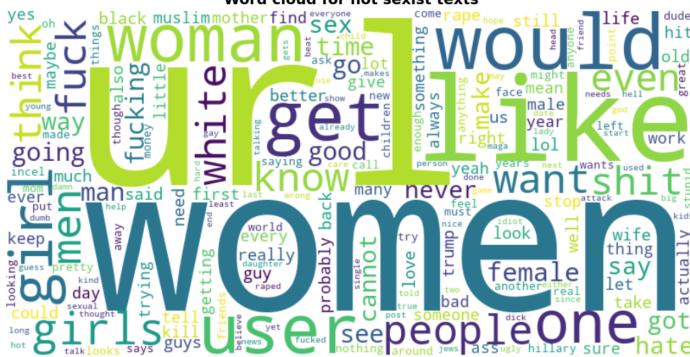
```
In [ ]: words = words.sort_values(by='freq_sexist', ascending=False)
    show_word_cloud(words['word'], words['freq_sexist'], 'Word cloud for sexist texts')
```



Word cloud of most frequent words of not sexist label

```
In [ ]: words = words.sort_values(by='freq_not_sexist', ascending=True)
    show_word_cloud(words['word'], words['freq_not_sexist'], 'Word cloud for not sexist texts')
```

Word cloud for not sexist texts



Emoji count

```
In [ ]: import emoji

words['is_emoji'] = words['word'].apply(lambda x: emoji.is_emoji(x))
 emoji_all_data = words[words['is_emoji'] == True].sort_values(by='freq', ascending=False)[['word emoji_all_data
```

Out[]:		word	freq	freq_sexist	freq_not_sexist
	0	\(\text{\tin}\text{\tetx{\texi}\text{\text{\text{\text{\text{\text{\text{\text{\text{\texi}\text{\text{\text{\text{\texi}\text{\text{\text{\text{\text{\texi}\text{\text{\text{\text{\text{\text{\text{\text{\text{\tetx{\texi}\text{\texi}\text{\texi}\text{\text{\texi}\text{\text{\ti}\text{\text{\text{\texit{\texi}\texit{\texi}\text{\text{\texi}\tittt{\texi}\texit{\texi}\text{\texi}\texit{\texi}\til\ti	174	46	128
	1	2	38	14	24
	2	• •	37	5	32
	3		22	5	17
	4	©	19	0	19
	•••				
	178	~	1	0	1
	179	<u>~</u>	1	0	1
	180	<u> </u>	1	0	1
	181	•	1	0	1
	182	2	1	0	1

183 rows × 4 columns

```
In [ ]: words[(words['is_emoji'] == True) & (words['freq_sexist'] > 0)].sort_values(by='freq_sexist', as
```

Out[]:		word	freq	freq_sexist	freq_not_sexist	is_emoji
	0	\(\text{\tin}\text{\tetx{\texi{\text{\texi}\text{\text{\text{\text{\text{\text{\text{\ti}\}\tittt{\text{\text{\texi}\text{\text{\text{\text{\texi{\tex{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\tetx{\ti}\tittt{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\ti}\tittit{\texi{\texi{\texi{\texi{\texi{\texi}\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi{\texi	174	46	128	True
	1	7	38	14	24	True
	2	•	8	5	3	True
	3	0.0	37	5	32	True
	4		22	5	17	True
	5	<u> </u>	15	5	10	True
	6	©	5	4	1	True
	7		5	3	2	True
	8	•	5	3	2	True
	9	•	8	3	5	True

```
In [ ]: emoji_all_data['text_ids'] = emoji_all_data['word'].apply(lambda x: get_text_ids(x))
    emoji_all_data['example_text'] = emoji_all_data['text_ids'].apply(lambda x: edos.loc[x[0], 'text
    emoji_all_data[['word', 'freq', 'freq_sexist', 'freq_not_sexist', 'example_text']].head(10)
```

example_text	freq_not_sexist	freq_sexist	freq	word	
i don't wanna suck any beef curtain flaps 😂	128	46	174	<u></u>	0
Aw, what's wrong nigger? You don't like your sheboon exposed for the savage she is? 🔣	24	14	38	3	1
Why does she have to be white? Do blackcels hate black females or something ? 😤	32	5	37	•••	2
He sure took her for a ride. 🤩	17	5	22		3
My mom does 😊	19	0	19	©	4
Lol sounds like me. I don't normally play the lottery but one yr my mom bought us all each 5 tickets for Christmas. I got nothing until my last ticket and it was a \$5 prize 😂 😂	15	2	17	\(\theta\)	5
Pray AlwaysusıL [USER] Or we'll rape your women and children? Migrant Urges Germans, Austrians To Give Refugees "Their Homes and Money" Zero Hedge [URL]	16	0	16	US	6
👍 YES she needs to and she won't shy away form it , more power to her , i like her spunk !! 🍨 🕰 🤗 😙 ❤	15	1	16	4	7
O come on there's no way any men are attracted to her shes a pig 🙄	10	5	15	•	8
Will be mentioning this to her next time we meet! Thanks twitter one can always find something good out of a bad situation eh 💍	14	0	14	8	9

```
In [ ]: print('Most frequent emoji and example text:')
   emoji_all_data.sort_values(by='freq', ascending=False)[['word', 'freq', 'example_text']].reset_i
```

Most frequent emoji and example text:

example_text	freq	word	
i don't wanna suck any beef curtain flaps 😂	174	~	0
Aw, what's wrong nigger? You don't like your sheboon exposed for the savage she is? 🤣	38	**	1
Why does she have to be white? Do blackcels hate black females or something?	37		2
He sure took her for a ride. 😂	22		3
My mom does 😊	19	0	4
ol sounds like me. I don't normally play the lottery but one yr my mom bought us all each 5 tickets for Christmas. I got nothing until my last ticket and it was a \$5 prize 😂 😂 😂	17	=	5
Pray Alwaysusıı [USER] Or we'll rape your women and children? Migrant Urges Germans, Austrians To Give Refugees "Their Homes and Money" Zero Hedge [URL]	16	US	6
YES she needs to and she won't shy away form it , more power to her , i like her spunk !! 🍨	16	Ġ	7
O come on there's no way any men are attracted to her shes a pig 🙄	15	•	8
oe mentioning this to her next time we meet! Thanks twitter one can always find something good out of a bad situation eh 🖔	14	8	9

```
In [ ]: print("Most frequent emoji in sexist texts and example text:")
        emoji_all_data.sort_values(by='freq_sexist', ascending=False)[['word', 'freq_sexist', 'example_to']
```

Most frequent emoji in sexist texts and example text:

'			
example_text	freq_sexist	word	
i don't wanna suck any beef curtain flaps 😂	46	~	0
Aw, what's wrong nigger? You don't like your sheboon exposed for the savage she is? 🤣	14	1	1
Why does she have to be white? Do blackcels hate black females or something ? $\stackrel{\textstyle \raisebox{.4ex}{$\scriptstyle \raisebox.4ex}{$\scriptstyle \raisebox{.4ex}{$\scriptstyle \raisebox{.4ex}{$\scriptstyle \raisebox{.4ex}{$\scriptstyle \raisebox{.4ex}{$\scriptstyle \raisebox.4ex}{$\scriptstyle \raisebox{.4ex}{$\scriptstyle \raisebox.4ex}{$\scriptstyle \raisebox.4ex}{}}}}}}}}}}}}}}}}}}}}}}$	5	•••	2
He sure took her for a ride. 😂	5	\odot	3
[USER] I am in love with her coat, such a beautiful colour - fun & modern cut, whilst staying classically elegant ♥	5	•	4
O come on there's no way any men are attracted to her shes a pig 🙄	5	~	5
Cardi b aint shit she should be in jail smh clown lookin ass bitch 🧐	4	©	5
layin pipe Imfao, 😂 😂 that pussy looking, milk toast mother fucker coudln't lay pipe if his name was Roto Rooter	3	~	7
"for unknown reasons" Lol I solved this one, the reason is because she was WHITE. Ok, another white targeted based on race, nothing to see here. Now get back on twitter and search for people telling the truth on islam	3	•	8
[USER] well, as there's no #finance tag I would leave it open if not, I'd pick #lifestyle> Money makes the world go round, so? Btw: we prefer #Brzz instead of BruXeLles	3	•	9

```
In [ ]: print("Most frequent emoji in not sexist texts and example text:")
        emoji_all_data.sort_values(by='freq_not_sexist', ascending=False)[['word', 'freq_not_sexist',
```

Most frequent emoji in not sexist texts and example text:

Out[]:		word	freq_not_sexist	example_text
	0		128	i don't wanna suck any beef curtain flaps 😂
	1		32	Why does she have to be white? Do blackcels hate black females or something?
	2	②	24	Aw, what's wrong nigger? You don't like your sheboon exposed for the savage she is?
	3	O	19	My mom does 😊
	4		17	He sure took her for a ride. 🤭
	5	US	16	Pray Alwaysusıl [USER] Or we'll rape your women and children? Migrant Urges Germans, Austrians To Give Refugees "Their Homes and Money" Zero Hedge [URL]
	6	=	15	Lol sounds like me. I don't normally play the lottery but one yr my mom bought us all each 5 tickets for Christmas. I got nothing until my last ticket and it was a \$5 prize \cong
	7	de la constant de la	15	👍 YES she needs to and she won't shy away form it , more power to her , i like her spunk !! 🎕 🕰 🤗 😙 ❤
	8	8	14	Will be mentioning this to her next time we meet! Thanks twitter one can always find something good out of a bad situation eh 🖔
	9	60	11	I despise the woman who wrote this article bashing Ivanka Trump. Please read this and send her a message via Twitter telling her how you hate it too. [URL] Amanda Carpenter [URL]

Save emoji dataframe to csv file

```
In [ ]: emoji_all_data.to_csv('emoji_all_data.csv', index=False)
```

CountVectorizer and chi2 score

CountVectorizer

In []: from sklearn.feature_selection import chi2

chi2_scores = chi2(X, np.array(edos['label_sexist']))

```
In [ ]: from sklearn.feature_extraction.text import CountVectorizer
         vectorizer = CountVectorizer(
             ngram_range=(1, 1), # trigram
             analyzer='word', # features made of words
             stop_words=list(stop_words_punct),
             lowercase=False,
             tokenizer=tokenize,
         X = vectorizer.fit_transform(edos['text'])
         vectorizer.get_feature_names_out()
       c:\Users\Minh Ha\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\feature_extract
       ion\text.py:528: UserWarning: The parameter 'token_pattern' will not be used since 'tokenizer' is
       not None'
         warnings.warn(
       c:\Users\Minh Ha\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\feature_extract
       ion\text.py:409: UserWarning: Your stop_words may be inconsistent with your preprocessing. Tokeniz
       ing the stop words generated tokens ['could', 'might', 'must', 'need', 'shall', 'would'] not in st
       op_words.
       warnings.warn(
\label{eq:out} {\tt Out[]: array(['##if', '#cw2', '(516)', ..., '$\frac{1}{2} \u200d$\text{$\Q$}', '$\frac{1}{2} \u200d$\text{$\Q$}'], }
               dtype=object)
In [ ]: X
Out[]: <20000x25155 sparse matrix of type '<class 'numpy.int64'>'
                 with 235423 stored elements in Compressed Sparse Row format>
         chi2 score
```

```
features = pd.DataFrame(columns=['word', 'chi2_score'])
features['word'] = vectorizer.get_feature_names_out()
features['chi2_score'] = chi2_scores[0]
features
```

```
Out[]:
                 word chi2_score
                  ##if
                         0.320481
                         0.320481
              1
                 #cw2
              2 (516)
                         0.320481
              3
                         3.120313
                    (8
                    /8
              4
                         0.320481
         25150
                    *
                         0.320481
         25151
                         0.320481
         25152
                   Q
                         0.640961
         25153
                   2
                         0.640961
         25154
                         0.640961
```

25155 rows \times 2 columns

```
In [ ]: features.shape, words.shape
```

Out[]: ((25155, 2), (24902, 5))

Out[]:

Merge 2 dataframes

```
In [ ]: # words.drop(columns=['chi2_scores'], inplace=True)
words = words.merge(features, on='word', how='outer')
words
```

	word	freq	freq_sexist	freq_not_sexist	is_emoji	chi2_score
0	pur	1.0	1.0	0.0	False	3.120313
1	dismemberment	1.0	1.0	0.0	False	3.120313
2	betabucks	1.0	1.0	0.0	False	3.120313
3	proverb	1.0	1.0	0.0	False	3.120313
4	braying	1.0	1.0	0.0	False	3.120313
•••						
25150	96	NaN	NaN	NaN	NaN	0.134079
25151	97	NaN	NaN	NaN	NaN	1.281923
25152	98	NaN	NaN	NaN	NaN	0.640961
25153	99	NaN	NaN	NaN	NaN	0.211401
25154	9966	NaN	NaN	NaN	NaN	0.320481

25155 rows \times 6 columns

```
In [ ]: words[words.isnull().any(axis=1)]
```

	word	freq	freq_sexist	freq_not_sexist	is_emoji	chi2_score
24902	0	NaN	NaN	NaN	NaN	6.229023
24903	00	NaN	NaN	NaN	NaN	0.640961
24904	000	NaN	NaN	NaN	NaN	1.281923
24905	00001	NaN	NaN	NaN	NaN	0.320481
24906	007	NaN	NaN	NaN	NaN	0.001160
•••						
25150	96	NaN	NaN	NaN	NaN	0.134079
25151	97	NaN	NaN	NaN	NaN	1.281923
25152	98	NaN	NaN	NaN	NaN	0.640961
25153	99	NaN	NaN	NaN	NaN	0.211401
25154	9966	NaN	NaN	NaN	NaN	0.320481

253 rows × 6 columns

Out[]:

```
In [ ]: words.dropna(inplace=True)
```

In []: words.sort_values(by='chi2_score', ascending=False).reset_index(drop=True).head(20)

Out[]:		word	freq	freq_sexist	freq_not_sexist	is_emoji	chi2_score
	0	bitch	591.0	577.0	14.0	False	1730.540692
	1	women	4334.0	1877.0	2457.0	False	854.725669
	2	pussy	250.0	224.0	26.0	False	580.533909
	3	whore	207.0	197.0	10.0	False	566.126961
	4	cunt	196.0	187.0	9.0	False	539.664231
	5	bitches	118.0	115.0	3.0	False	343.890048
	6	men	1464.0	658.0	806.0	False	340.493873
	7	slut	123.0	109.0	14.0	False	277.099631
	8	whores	79.0	75.0	4.0	False	214.644166
	9	thots	58.0	57.0	1.0	False	172.831343
	10	url	3414.0	507.0	2907.0	False	164.805184
	11	cunts	55.0	52.0	3.0	False	147.785656
	12	sluts	50.0	43.0	7.0	False	103.663251
	13	feminists	118.0	75.0	43.0	False	99.104493
	14	hag	40.0	35.0	5.0	False	87.009890
	15	hoe	29.0	28.0	1.0	False	82.436068
	16	attention	131.0	76.0	55.0	False	81.163227
	17	pussies	27.0	26.0	1.0	False	76.209339
	18	thot	27.0	26.0	1.0	False	76.209339
	19	twat	21.0	21.0	0.0	False	65.526576

Word length

```
In [ ]: words['len'] = words['word'].apply(lambda x: len(x))
    words.sort_values(by='len', ascending=False).reset_index(drop=True).head(20)
```

	word	freq	freq_sexist	freq_not_sexist	is_emoji	chi2_score	len
0	000000000000000000000000000000000000000	1.0	0.0	1.0	False	0.320481	24
1	raaaaaavvvvvvinnnnnggggg	1.0	0.0	1.0	False	0.320481	24
2	hahaahahhhhahahahhhaa	1.0	0.0	1.0	False	0.320481	21
3	hahahahahahahahaha	2.0	2.0	0.0	False	6.240626	20
4	mischaracterization	1.0	1.0	0.0	False	3.120313	19
5	alkjtlkjafdlkdaflkj	1.0	0.0	1.0	False	0.320481	19
6	oversimplification	1.0	0.0	1.0	False	0.320481	18
7	ssssssssshadilay	1.0	1.0	0.0	False	3.120313	18
8	feeeeeeeeeelings	1.0	1.0	0.0	False	3.120313	18
9	disproportionately	1.0	0.0	1.0	False	0.320481	18
10	unfollowtrump4prez	1.0	0.0	1.0	False	0.320481	18
11	indistinguishable	2.0	1.0	1.0	False	0.720397	17
12	institutionalized	1.0	0.0	1.0	False	0.320481	17
13	opportunistically	1.0	1.0	0.0	False	3.120313	17
14	reeeeeeeeeeee	1.0	0.0	1.0	False	0.320481	17
15	characterizations	1.0	0.0	1.0	False	0.320481	17
16	disproportionate	1.0	1.0	0.0	False	3.120313	16
17	unapologetically	3.0	0.0	3.0	False	0.961442	16
18	responsibilities	6.0	4.0	2.0	False	5.867822	16
19	characterization	1.0	0.0	1.0	False	0.320481	16

```
In [ ]: words['len'].describe()
```

```
Out[]: count
                 24902.000000
        mean
                     6.930287
                     2.470459
        std
                     1.000000
        min
        25%
                     5.000000
        50%
                     7.000000
        75%
                     9.000000
                    24.000000
        max
        Name: len, dtype: float64
```

Out[]:

Out[]:

Correlation between word length, frequency and chi2 score

```
In [ ]: words[['freq', 'freq_sexist', 'freq_not_sexist', 'chi2_score', 'len']].corr()
```

	freq	freq_sexist	freq_not_sexist	chi2_score	len
freq	1.000000	0.933274	0.987631	0.323695	-0.094555
freq_sexist	0.933274	1.000000	0.865414	0.530709	-0.079310
freq_not_sexist	0.987631	0.865414	1.000000	0.219888	-0.097286
chi2_score	0.323695	0.530709	0.219888	1.000000	-0.021536
len	-0.094555	-0.079310	-0.097286	-0.021536	1.000000

Words not in English dictionary

Select words not in English dictionary

```
In [ ]: words['found'] = words['word'].apply(lambda x: x in english_words)
words[(words['found'] == False) & (words['is_emoji'] == False)].reset_index(drop=True)
```

	word	freq	freq_sexist	freq_not_sexist	is_emoji	chi2_score	len	found
0	betabucks	1.0	1.0	0.0	False	3.120313	9	False
1	creampied	1.0	1.0	0.0	False	3.120313	9	False
2	muthafuck	1.0	1.0	0.0	False	3.120313	9	False
3	mostt	1.0	1.0	0.0	False	3.120313	5	False
4	ininity	1.0	1.0	0.0	False	3.120313	7	False
•••								
2559	wife's	37.0	7.0	30.0	False	0.576431	6	False
2560	cucks	46.0	15.0	31.0	False	1.740269	5	False
2561	woman's	61.0	18.0	43.0	False	0.910659	7	False
2562	women's	120.0	45.0	75.0	False	11.427815	7	False
2563	gabfam	118.0	15.0	103.0	False	8.576693	6	False

2564 rows × 8 columns

Out[]:

Out[]:

```
In []: for word in words[(words['found'] == False) & (words['is_emoji'] == False)]['word']:
    # if word not contains any alphabetic character, delete it
    if not any(c.isalpha() for c in word):
        words.drop(words[words['word'] == word].index, inplace=True)
        continue
    # if word ends with "'s" and the word without "'s" is in english words, delete it
    if word.endswith("'s") and word[:-2] in english_words:
        words.drop(words[words['word'] == word].index, inplace=True)
        continue
words[(words['found'] == False) & (words['is_emoji'] == False)].reset_index(drop=True)
```

	word	freq	freq_sexist	freq_not_sexist	is_emoji	chi2_score	len	found
0	betabucks	1.0	1.0	0.0	False	3.120313	9	False
1	creampied	1.0	1.0	0.0	False	3.120313	9	False
2	muthafuck	1.0	1.0	0.0	False	3.120313	9	False
3	mostt	1.0	1.0	0.0	False	3.120313	5	False
4	ininity	1.0	1.0	0.0	False	3.120313	7	False
•••								
1953	ausfam	25.0	4.0	21.0	False	0.930279	6	False
1954	libtards	25.0	3.0	22.0	False	2.047818	8	False
1955	:d	28.0	0.0	28.0	False	8.973458	2	False
1956	cucks	46.0	15.0	31.0	False	1.740269	5	False
1957	gabfam	118.0	15.0	103.0	False	8.576693	6	False

1958 rows × 8 columns

```
In [ ]: not_found_words = words[(words['found'] == False) & (words['is_emoji'] == False)][['word', 'freq
not_found_words
```

	word	freq	freq_sexist	freq_not_sexist	chi2_score
0	betabucks	1.0	1.0	0.0	3.120313
1	creampied	1.0	1.0	0.0	3.120313
2	muthafuck	1.0	1.0	0.0	3.120313
3	mostt	1.0	1.0	0.0	3.120313
4	ininity	1.0	1.0	0.0	3.120313
•••					
1953	ausfam	25.0	4.0	21.0	0.930279
1954	libtards	25.0	3.0	22.0	2.047818
1955	:d	28.0	0.0	28.0	8.973458
1956	cucks	46.0	15.0	31.0	1.740269
1957	gabfam	118.0	15.0	103.0	8.576693

1958 rows × 5 columns

Out[]:

Get 1 example text containing words not in English dictionary and get all texts' id containing words not in English dictionary

```
In [ ]: not_found_words['text_ids'] = not_found_words['word'].apply(lambda x: get_text_ids(x))
not_found_words['example_text'] = not_found_words['text_ids'].apply(lambda x: edos.loc[x[0]]['text_not_found_words[['word', 'freq', 'chi2_score', 'example_text']].reset_index(drop=True)
```

Out[]:		word	freq	chi2_score	example_text
	0	betabucks	1.0	3.120313	aww you could have pumped and dumped her. at least you took the 2nd best option, anything but betabucks
	1	creampied	1.0	3.120313	Yeah, my wife has 7 kids by 9 different men, got creampied by 50, spitroasted by 12 pairs, and is in an open marriage with me, but she comes home to me every night. So in the end, who wins?
	2	muthafuck	1.0	3.120313	im wiggapriveledge yo git ti fuck oota here yo niggapriveledge muthafuck ya on ma turf ya all pay me ho.
	3	mostt	1.0	3.120313	In my experience women don't recommend guys to their friends. If she doesn't want him then mostt likely other women don't
	4	ininity	1.0	3.120313	trousers to ininity and beyond, people post it if they think it haven't been posted before, while it is annoying, i do still see a point of re-posting this kind of post, so every new red pilled man can see how annoying/toxic female nature is
	•••	•••			
	1953	ausfam	25.0	0.930279	Schoolgirl, 15, has her jaw broken in three places during vicious brawl that spilled onto the road and saw students assault parents [URL] #AusFam #AusPol #ausfam #auspol #Ausfam #Auspol #Australia
	1954	libtards	25.0	2.047818	#EmpressMichelle just needs to pack her shit, #stfu and move to Venezuela. Thank you to all the #Libtards who gave us these two pieces of shit [URL]
	1955	:d	28.0	8.973458	TL:DR unclear covered her in shit and now she is calling the police
	1956	cucks	46.0	1.740269	AITA is full of cucks who downvote anything where a man is in the right against a woman.
	1957	gabfam	118.0	8.576693	Remember when the German lady was kicked from behind down the stairs? Well here's the scum that did it [URL] #MEGA #SpeakFreely #GabFam

1958 rows × 4 columns