DATA IMPORT

df = pd.read_csv("Womens Clothing E-Commerce Reviews.csv",index_col=0)
df.head()

	Clothing ID	Age	Title	Review Text	Rating	Recommended IND	Positive Feedback Count	Division Name	Departme Na
(0 767	33	NaN	Absolutely wonderful - silky and sexy and comf	4	1	0	Initmates	Intima
4									•

```
df.drop(labels=['Title','Clothing ID'],axis=1,inplace=True)

df.head()
```

	Age	Review Text	Rating	Recommended IND	Positive Feedback Count	Division Name	Department Name	Class Name
	0 33	Absolutely wonderful - silky and sexy and	4	1	0	Initmates	Intimate	Intimates
df.is	snull().sum	1()						
df.dı	Division N Department Class Name dtype: int	ed IND Feedback Count Name Name	0 845 0 0 14 14 14	sion Name'],in	place=True)			
df.is	snull().sum	1()						
	Age Review Tex Rating Recommende Positive F Division N Department Class Name dtype: int	ed IND Feedback Count Name Name	0 0 0 0 0 0					

' '.join(df["Review Text"].tolist())

'Absolutely wonderful - silky and sexy and comfortable Love this dress! it\'s sooo pre tty. i happened to find it in a store, and i\'m glad i did bc i never would have order ed it online bc it\'s petite. i bought a petite and am 5\'8". i love the length on me - hits just a little below the knee. would definitely be a true midi on someone who is truly petite. I had such high hopes for this dress and really wanted it to work for me. i initially ordered the petite small (my usual size) but i found this to be outrageousl v small. so small in fact that i could not zip it up! i reordered it in petite medium.

Text Cleaning

```
contractions = {
   "ain't":"am not",
   "aren't":"are not",
```

```
"can't": "cannot",
"can't've": "cannot have",
"'cause": "because",
"could've": "could have",
"couldn't": "could not",
"couldn't've": "could not have",
"didn't": "did not",
"don't": "do not",
"doesn't": "does not",
"hadn't": "had not",
"hasn't": "has not",
"he'd": "he would",
"he'd've": "he would have",
"he'll": "he will",
"he'll've": "he will have",
"he's": "he is",
"how'd": "how did",
"how'd'y": "how do you",
"how'll": "how will",
"how's": "how does",
"i'll":"i will",
"i'd":"i would",
"i'd've":"i would have",
"i'll":"i will",
"i'll've":"i will have",
"i'm":"i am",
"i've":"i have",
"isn't":"is not",
"it'd":"it would",
"it'd've": "it would have",
"it'll":"it will",
"it'll've": "it will have",
"it's":"it is",
"let's":"let us",
"ma'am": "madam",
"mayn't": "may not",
"might've": "might have",
"mightn't've": "might not have",
"must've": "must not",
"mustn't": "must not",
"mustn't've": "must not have",
"need't": "need not",
"needn't've": "need not have",
"o'clock": "of the clock",
"oughtn't": "ought not",
"oughtn't've": "ought not have",
"shan't": "shall not",
"sha'n't": "shall not",
"shan't've": "shall not have",
"she'd": "she would",
```

"she'd've": "she would have",

```
"she'll": "she will",
    "she'll've": "she will have",
    "she's": "she is",
    "should've": "should have",
    "shouldn't": "should not",
    "shouldn't've": "should not have",
    "so've": "so have",
    "so's":"so is",
    "that's": "that is",
    "that'd": "that would",
    "that'd've": "that would have",
    "there'd": "there would",
    "there'd've": "there would have",
    "there's": "there is",
    "they'd": "they would",
    "they'd've": "they would have",
    "they'll":"they will",
    "they'll've": "they will have",
    "they're": "they are",
    "they've": "they have",
    "to've": "to have",
    "wasn't":"was not",
    " u ":" you ",
    " ur ":" your ",
    " n ":" and "
}
def cont_to_exp(x):
  if type(x) is str:
   x = x.replace('\\','')
    for key in contractions:
      value = contractions[key]
      x = x.replace(key,value)
    return x
  else:
    return x
x = "i don't know what it's going to cost. You should've told me the cost\""
print(cont_to_exp(x))
     i do not know what it is going to cost. You should have told me the cost"
df['Review Text'] = df['Review Text'].apply(lambda x: cont_to_exp(x))
df.head()
```

		Age Review Text		Review Text Rating		Recommended IND Positive Feedback Count		Department Name	Class Name	
	0	33	Absolutely wonderful - silky and sexy and comf	4	1	0	Initmates	Intimate	Intimates	
	1	34	Love this dress! it is sooo pretty. i	5	1	4	General	Dresses	Dresses	
print	('	'.joi	n(df['Review Text	:'].tolis	st())[:1000])					

Absolutely wonderful - silky and sexy and comfortable Love this dress! it is sooo prett

Feature Engineering

df.head()

	Age	Review Text	Review Text Rating		Positive Feedback Count	Division Name	Department Name	Class Name	
0	33	Absolutely wonderful - silky and sexy and comf	4	1	0	Initmates	Intimate	Intimates	
1	34	Love this dress! it is sooo pretty. i	5	1	4	General	Dresses	Dresses	

from textblob import TextBlob

```
df['polarity'] = df['Review Text'].apply(lambda x: TextBlob(x).sentiment.polarity)

#pip install textblob

df['review_length'] = df['Review Text'].apply(lambda x:len(x))

df['word_count'] = df['Review Text'].apply(lambda x:len(x.split()))

def get_avg_word_len(x):
    words = x.split()
    word len = 0
```

```
for word in words:
    word_len = word_len + len(word)

return word_len/len(words)

df['avg_word_len'] = df['Review Text'].apply(lambda x: get_avg_word_len(x))

df.head()
```

	Age	Review Text	Rating	Recommended IND	Positive Feedback Count	Division Name	Department Name	Class Name	polarit
0	33	Absolutely wonderful - silky and sexy and comf	4	1	0	Initmates	Intimate	Intimates	0.63333
1	34	Love this dress! it is sooo pretty. i happen	5	1	4	General	Dresses	Dresses	0.33958
4									>

Distribution of Sentiment Polarity

df.iplot()

- Distribution of Reviews Rating and Reviewers Age

df['Rating'].iplot(kind='hist',xTitle='Rating',yTitle='Count',title='Review Rating Distributi

Distribution of Review Text Length and Word Length

Distribution of Department, Division and Class

```
df['Department Name'].value_counts()
```

Tops	10048
Dresses	6145
Bottoms	3662
Intimate	1653
Jackets	1002
Trend	118

Name: Department Name, dtype: int64

df.groupby('Department Name').count()

		Age	Review Text	Rating	Recommended IND	Positive Feedback Count	Division Name	Class Name	polarity	reı
ı	Department Name									
	Bottoms	3662	3662	3662	3662	3662	3662	3662	3662	
	Dresses	6145	6145	6145	6145	6145	6145	6145	6145	
	Intimate	1653	1653	1653	1653	1653	1653	1653	1653	
	Jackets	1002	1002	1002	1002	1002	1002	1002	1002	
4										•

df['Department Name'].value_counts().iplot(kind="bar",yTitle="Count",xTitle='Department')

→ Distribution of Unigram, Bigram and Trigram

→ Unigram

```
vec = CountVectorizer().fit(x)
bow = vec.transform(x)
sum words = bow.sum(axis=0)
words freq = [(word, sum words[0, idx]) for word, idx in vec.vocabulary .items()]
words freq = sorted(words freq, key = lambda x: x[1], reverse = True)
words_freq[:2]
     [('this', 2), ('list', 2)]
sum_words
     matrix([[1, 2, 1, 2]])
def get top n words(x,n):
    vec = CountVectorizer().fit(x)
    bow = vec.transform(x)
    sum words = bow.sum(axis=0)
    words_freq = [(word, sum_words[0, idx]) for word, idx in vec.vocabulary_.items()]
    words freq = sorted(words freq, key = lambda x: x[1], reverse = True)
    return words freq[:n]
get_top_n_words(x,2)
     [('this', 2), ('list', 2)]
words = get top n words(df['Review Text'],20)
words
     [('the', 76159),
      ('it', 49273),
      ('and', 49008),
      ('is', 38103),
      ('this', 25758),
      ('to', 24577),
      ('in', 20722),
      ('but', 16549),
      ('not', 16203),
      ('on', 15325),
      ('for', 13994),
      ('of', 13428),
      ('was', 12923),
      ('with', 12797),
      ('so', 12017),
```

```
('am', 11174),
    ('my', 11027),
    ('dress', 10567),
    ('that', 10011),
    ('love', 8945)]

df1 = pd.DataFrame(words, columns=['Unigram', 'Frequency'])
df1 = df1.set_index('Unigram')
df1.iplot(kind="bar",xTitle='Unigram',yTitle='Count',title='Top 20 Unigram Words')
```

→ Bigram

```
vec = CountVectorizer(ngram_range=(2,2)).fit(x)
bow = vec.transform(x)
sum_words = bow.sum(axis=0)
words_freq = [(word, sum_words[0, idx]) for word, idx in vec.vocabulary_.items()]
words_freq = sorted(words_freq, key = lambda x: x[1], reverse = True)
words_freq[:2]
```

```
[('this is', 1), ('is the', 1)]
def get top n words(x,n):
    vec = CountVectorizer(ngram_range=(2,2)).fit(x)
    bow = vec.transform(x)
    sum words = bow.sum(axis=0)
    words_freq = [(word, sum_words[0, idx]) for word, idx in vec.vocabulary_.items()]
    words freq = sorted(words freq, key = lambda x: x[1], reverse = True)
    return words freq[:n]
get_top_n_words(x,2)
     [('this is', 1), ('is the', 1)]
words = get_top_n_words(df['Review Text'],20)
words
     [('it is', 12525),
      ('in the', 7169),
      ('and the', 5613),
      ('this dress', 4758),
      ('on the', 4337),
      ('of the', 3933),
      ('and it', 3723),
      ('it was', 3287),
      ('this top', 2939),
      ('this is', 2772),
      ('is very', 2729),
      ('the fabric', 2655),
      ('love the', 2641),
      ('did not', 2486),
      ('love this', 2331),
      ('to wear', 2266),
      ('but it', 2196),
      ('the dress', 2180),
      ('is not', 2130),
      ('do not', 2117)]
df2 = pd.DataFrame(words, columns=['Bigram', 'Frequency'])
df2 = df2.set index('Bigram')
df2.iplot(kind="bar",xTitle='Bigram',yTitle='Count',title='Top 20 Bigram Words')
```

→ Trigram

```
def get_top_n_words(x,n):
    vec = CountVectorizer(ngram_range=(3,3)).fit(x)
    bow = vec.transform(x)
    sum words = bow.sum(axis=0)
    words_freq = [(word, sum_words[0, idx]) for word, idx in vec.vocabulary_.items()]
    words freq = sorted(words freq, key = lambda x: x[1], reverse = True)
    return words freq[:n]
get_top_n_words(x,3)
     [('this is the', 1), ('is the list', 1), ('the list list', 1)]
words = get_top_n_words(df['Review Text'],20)
words
     [('true to size', 1316),
      ('the fabric is', 1301),
      ('and it is', 1124),
      ('this dress is', 1123),
      ('it is very', 976),
      ('but it is', 921),
      ('it is not', 910),
```

```
('the material is', 896),
      ('in the store', 728),
      ('on the model', 725),
      ('the fit is', 696),
      ('this top is', 672),
      ('the color is', 630),
      ('love this dress', 605),
      ('it is so', 584),
      ('tried it on', 574),
      ('and it was', 550),
      ('in the back', 528),
      ('the dress is', 518),
      ('up or down', 518)]
df3 = pd.DataFrame(words, columns=['Trigram', 'Frequency'])
df3 = df3.set_index('Trigram')
df3.iplot(kind="bar",xTitle='Trigram',yTitle='Count',title='Top 20 Trigram Words')
```

Distribution of Unigram, Bigram and Trigram without STOPWORDS

Unigram

```
def get_top_n_words(x,n):
    vec = CountVectorizer(ngram range=(1,1),stop words='english').fit(x)
    bow = vec.transform(x)
    sum words = bow.sum(axis=0)
    words_freq = [(word, sum_words[0, idx]) for word, idx in vec.vocabulary_.items()]
    words_freq = sorted(words_freq, key = lambda x: x[1], reverse = True)
    return words freq[:n]
get_top_n_words(x,3)
     [('list', 2)]
words = get_top_n_words(df['Review Text'],20)
words
     [('dress', 10567),
      ('love', 8945),
      ('size', 8770),
      ('fit', 7322),
      ('like', 7148),
      ('wear', 6434),
      ('great', 6114),
      ('just', 5604),
      ('fabric', 4797),
      ('small', 4726),
      ('color', 4604),
      ('look', 4039),
      ('really', 3924),
      ('ordered', 3850),
      ('little', 3773),
      ('perfect', 3772),
      ('flattering', 3519),
      ('did', 3447),
      ('soft', 3336),
      ('comfortable', 3058)]
df2 = pd.DataFrame(words, columns=['Unigram', 'Frequency'])
df2 = df2.set index('Unigram')
df2.iplot(kind="bar",xTitle='Unigram',yTitle='Count',title='Top 20 Unigram Words')
```

→ Bigram

```
def get_top_n_words(x,n):
    vec = CountVectorizer(ngram_range=(2,2),stop_words='english').fit(x)
    bow = vec.transform(x)
    sum_words = bow.sum(axis=0)
    words_freq = [(word, sum_words[0, idx]) for word, idx in vec.vocabulary_.items()]
    words_freq = sorted(words_freq, key = lambda x: x[1], reverse = True)
    return words_freq[:n]

get_top_n_words(x,2)
    [('list list', 1)]

words = get_top_n_words(df['Review Text'],20)
words

[('true size', 1347),
    ('love dress', 766),
    ('usually wear', 694),
    ('looks great', 620),
```

```
('fit perfectly', 609),
      ('size small', 540),
      ('fits perfectly', 489),
      ('usual size', 479),
      ('just right', 434),
      ('look like', 426),
      ('ordered size', 375),
      ('love love', 374),
      ('looks like', 373),
      ('runs large', 367),
      ('super cute', 363),
      ('highly recommend', 363),
      ('wear size', 361),
      ('fabric soft', 356),
      ('feel like', 349),
      ('fit great', 348)]
df2 = pd.DataFrame(words, columns=['Bigram','Frequency'])
df2 = df2.set index('Bigram')
df2.iplot(kind="bar",xTitle='Unigram',yTitle='Count',title='Top 20 Bigram Words')
```

Trigram

```
def get_top_n_words(x,n):
    vec = CountVectorizer(ngram range=(3,3),stop words='english').fit(x)
    bow = vec.transform(x)
    sum words = bow.sum(axis=0)
    words freq = [(word, sum words[0, idx]) for word, idx in vec.vocabulary .items()]
    words_freq = sorted(words_freq, key = lambda x: x[1], reverse = True)
    return words freq[:n]
#get top n words(x,3)
words = get top n words(df['Review Text'],20)
words
     [('fits true size', 280),
      ('fit true size', 229),
      ('runs true size', 148),
      ('love love love', 143),
      ('usually wear size', 136),
      ('ordered usual size', 107),
      ('does run large', 96),
      ('really wanted love', 94),
      ('wanted love dress', 88),
      ('usually wear small', 80),
      ('small fit perfectly', 77),
      ('just did work', 71),
      ('looks great jeans', 70),
      ('fit like glove', 69),
      ('usually wear medium', 68),
      ('normally wear size', 68),
      ('small fits perfectly', 65),
      ('size fit perfectly', 65),
      ('fits like glove', 65),
      ('usual size small', 64)]
df2 = pd.DataFrame(words, columns=['Trigram', 'Frequency'])
df2 = df2.set index('Trigram')
df2.iplot(kind="bar",xTitle='Trigram',yTitle='Count',title='Top 20 Trigram Words')
```

Distribution of Top 20 Parts-of-Speech POS tags

```
import nltk
nltk.download('punkt')
nltk.download('averaged perceptron tagger')
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk data]
                   Unzipping tokenizers/punkt.zip.
     [nltk_data] Downloading package averaged_perceptron_tagger to
     [nltk data]
                     /root/nltk data...
                   Unzipping taggers/averaged perceptron tagger.zip.
     [nltk data]
     True
print(str(df['Review Text']))
              Absolutely wonderful - silky and sexy and comf...
              Love this dress! it is sooo pretty. i happen...
     1
              I had such high hopes for this dress and reall...
              I love, love, love this jumpsuit. it is fun, f...
              This shirt is very flattering to all due to th...
     23481
              I was very happy to snag this dress at such a ...
     23482
              It reminds me of maternity clothes. soft, stre...
              This fit well, but the top was very see throug...
     23483
     23484
              I bought this dress for a wedding i have this ...
              This dress in a lovely platinum is feminine an...
     23485
     Name: Review Text, Length: 22628, dtype: object
```

```
blob = TextBlob(str(df['Review Text']))
```

blob

```
TextBlob("0
                   Absolutely wonderful - silky and sexy and comf...
         Love this dress! it is sooo pretty. i happen...
1
2
         I had such high hopes for this dress and reall...
3
         I love, love, love this jumpsuit. it is fun, f...
4
         This shirt is very flattering to all due to th...
         I was very happy to snag this dress at such a ...
23481
         It reminds me of maternity clothes. soft, stre...
23482
23483
         This fit well, but the top was very see throug...
         I bought this dress for a wedding i have this ...
23484
         This dress in a lovely platinum is feminine an...
23485
Name: Review Text, Length: 22628, dtype: object")
```

blob.tags

```
( was , עטע )
('very', 'RB'),
('happy', 'JJ'),
('to', 'TO'),
('snag', 'VB'),
('this', 'DT'),
('dress', 'NN'),
('at', 'IN'),
('such', 'PDT'),
('a', 'DT'),
('23482', 'CD'),
('It', 'PRP'),
('reminds', 'VBZ'),
('me', 'PRP'),
('of', 'IN'),
('maternity', 'NN'),
('clothes', 'NNS'),
('soft', 'JJ'),
('stre', 'NN'),
('23483', 'CD'),
('This', 'DT'),
('fit', 'NN'),
('well', 'RB'),
('but', 'CC'),
('the', 'DT'),
('top', 'NN'),
('was', 'VBD'),
('very', 'RB'),
('see', 'JJ'),
('throug', 'NN'),
('23484', 'CD'),
('I', 'PRP'),
('bought', 'VBD'),
('this', 'DT'),
('dress', 'NN'),
('for', 'IN'),
('a'
      'DT')
```

```
( u , レ i / )
      ('wedding', 'NN'),
      ('i', 'NN'),
      ('have', 'VBP'),
      ('this', 'DT'),
      ('23485', 'CD'),
      ('This', 'DT'),
      ('dress', 'NN'),
      ('in', 'IN'),
      ('a', 'DT'),
      ('lovely', 'JJ'),
      ('platinum', 'NN'),
      ('is', 'VBZ'),
      ('feminine', 'JJ'),
      ('an', 'DT'),
      ('Name', 'NN'),
      ('Review', 'NNP'),
      ('Text', 'NNP'),
      ('Length', 'NNP'),
      ('22628', 'CD'),
      ('dtype', 'NN'),
      ('object', 'NN')]
nltk.download('tagsets')
     [nltk data] Downloading package tagsets to /root/nltk data...
     [nltk data] Unzipping help/tagsets.zip.
     True
print(nltk.help.upenn tagset())
         occasionally anabacingly madaeningly advendinosity professeaty
         stirringly prominently technologically magisterially predominately
         swiftly fiscally pitilessly ...
     RBR: adverb, comparative
         further gloomier grander graver greater grimmer harder harsher
         healthier heavier higher however larger later leaner lengthier less-
         perfectly lesser lonelier longer louder lower more ...
     RBS: adverb, superlative
         best biggest bluntest earliest farthest first furthest hardest
         heartiest highest largest least less most nearest second tightest worst
     RP: particle
         aboard about across along apart around aside at away back before behind
         by crop down ever fast for forth from go high i.e. in into just later
         low more off on open out over per pie raising start teeth that through
         under unto up up-pp upon whole with you
     SYM: symbol
         % & ' '' ''. ) ). * + ,. < = > @ A[fj] U.S U.S.S.R * ** ***
     TO: "to" as preposition or infinitive marker
         to
     UH: interjection
         Goodbye Goody Gosh Wow Jeepers Jee-sus Hubba Hey Kee-reist Oops amen
         huh howdy uh dammit whammo shucks heck anyways whodunnit honey golly
         man baby diddle hush sonuvabitch ...
```

```
8/24/22, 1:27 AM
```

```
EDA on Women's E commerce Clothing Reviews (1).ipynb - Colaboratory
     vs: verb, base torm
         ask assemble assess assign assume atone attention avoid bake balkanize
         bank begin behold believe bend benefit bevel beware bless boil bomb
         boost brace break bring broil brush build ...
     VBD: verb, past tense
         dipped pleaded swiped regummed soaked tidied convened halted registered
         cushioned exacted snubbed strode aimed adopted belied figgered
         speculated wore appreciated contemplated ...
     VBG: verb, present participle or gerund
         telegraphing stirring focusing angering judging stalling lactating
         hankerin' alleging veering capping approaching traveling besieging
         encrypting interrupting erasing wincing ...
     VBN: verb, past participle
         multihulled dilapidated aerosolized chaired languished panelized used
         experimented flourished imitated reunifed factored condensed sheared
         unsettled primed dubbed desired ...
     VBP: verb, present tense, not 3rd person singular
         predominate wrap resort sue twist spill cure lengthen brush terminate
         appear tend stray glisten obtain comprise detest tease attract
         emphasize mold postpone sever return wag ...
     VBZ: verb, present tense, 3rd person singular
         bases reconstructs marks mixes displeases seals carps weaves snatches
         slumps stretches authorizes smolders pictures emerges stockpiles
         seduces fizzes uses bolsters slaps speaks pleads ...
     WDT: WH-determiner
         that what whatever which whichever
     WP: WH-pronoun
         that what whatever whatsoever which who whom whosoever
     WP$: WH-pronoun, possessive
         whose
     WRB: Wh-adverb
         how however whence whenever where whereby whereever wherein whereof why
     ``: opening quotation mark
     None
pos df = pd.DataFrame(blob.tags, columns=['words','pos'])
pos df = pos df['pos'].value counts()
pos df
     NN
            23
     DT
            15
     JJ
            13
     CD
            11
```

```
PRP
         8
RB
         6
         5
VBZ
VBP
         5
         5
IN
CC
         4
NNP
         4
VBD
         4
TO
         3
         2
NNS
```

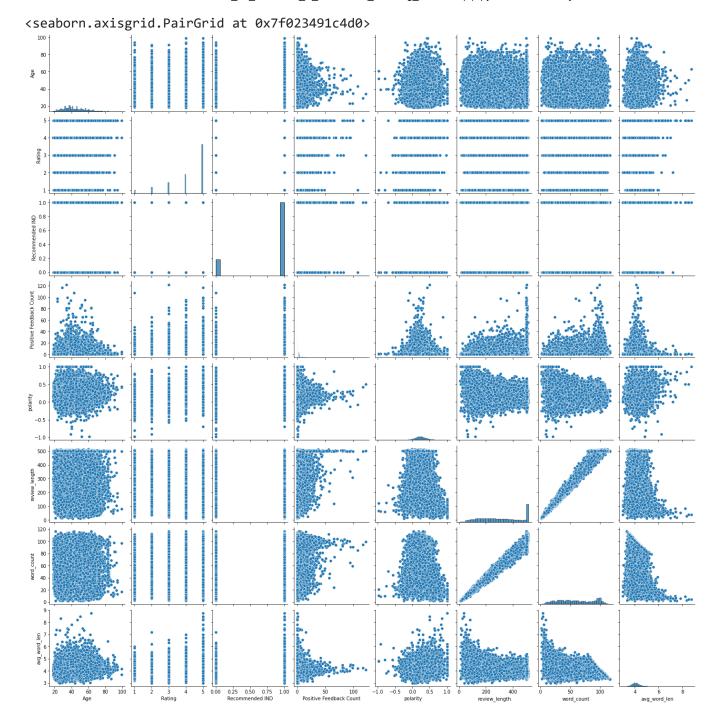
PDT 1

Name: pos, dtype: int64

pos_df.iplot(kind='bar')

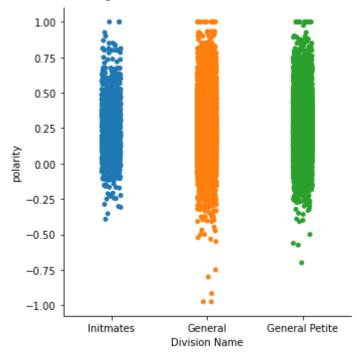
→ Bivariate Analysis

sns.pairplot(df)



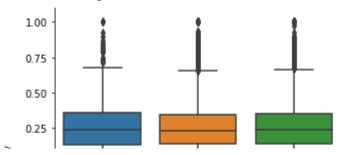
sns.catplot(x='Division Name',y='polarity',data=df)





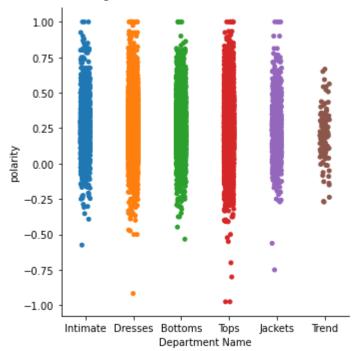
sns.catplot(x='Division Name',y='polarity',data=df,kind='box')

<seaborn.axisgrid.FacetGrid at 0x7f0233e88e50>



sns.catplot(x='Department Name',y='polarity',data=df)

<seaborn.axisgrid.FacetGrid at 0x7f0231bce850>



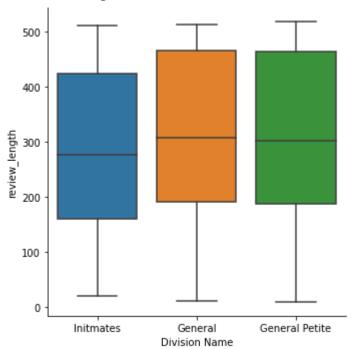
sns.catplot(x='Department Name',y='polarity',data=df,kind='box')

<seaborn.axisgrid.FacetGrid at 0x7f0231613710>



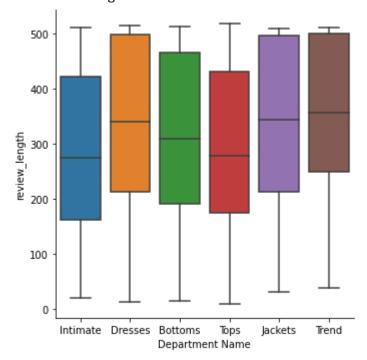
sns.catplot(x='Division Name',y="review_length",data=df,kind="box")

<seaborn.axisgrid.FacetGrid at 0x7f0231499e50>



sns.catplot(x='Department Name',y="review_length",data=df,kind="box")

<seaborn.axisgrid.FacetGrid at 0x7f02319851d0>



Distribution of Sentiment Polarity with Recommended Scores

```
import plotly.express as px
import plotly.graph_objects as go

x1 = df[df['Recommended IND']==1]['polarity']
x0=df[df['Recommended IND']==0]['polarity']

trace0 = go.Histogram(x=x0,name='Not Recommended',opacity=0.7)

trace1 = go.Histogram(x=x1,name='Recommended',opacity=0.7)

data = [trace0,trace1]
layout = go.Layout(barmode='overlay',
   title='Distribution of sentiment polarity based on the Recommended Scores')
fig = go.Figure(data=data,layout=layout)
iplot(fig)
```

Distribution of Ratings Based on the Recommendation

```
x1 = df[df['Recommended IND']==1]['Rating']
x0=df[df['Recommended IND']==0]['Rating']

trace0 = go.Histogram(x=x0,name='Not Recommended',opacity=0.7)

trace1 = go.Histogram(x=x1,name='Recommended',opacity=0.7)

data = [trace0,trace1]
layout = go.Layout(barmode='overlay',
   title='Distribution of Rating Reviews based on the Recommended Scores')
fig = go.Figure(data=data,layout=layout)
iplot(fig)
```

```
x1 = df[df['Recommended IND']==1]['polarity']
x0=df[df['Recommended IND']==0]['polarity']

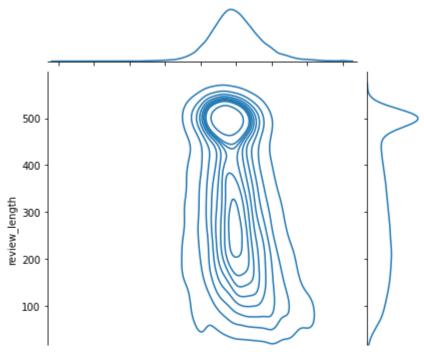
trace0 = go.Histogram(x=x0,name='Not Recommended',opacity=0.7)

trace1 = go.Histogram(x=x1,name='Recommended',opacity=0.7)

data = [trace0,trace1]
layout = go.Layout(barmode='overlay',
   title='Distribution of sentiment polarity based on the Recommended Scores')
fig = go.Figure(data=data,layout=layout)
iplot(fig)
```

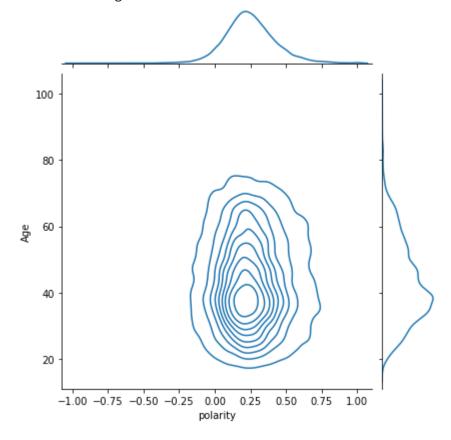
```
sns.jointplot(x='polarity',y='review length',data=df,kind='kde')
```

<seaborn.axisgrid.JointGrid at 0x7f0230ce4590>



sns.jointplot(x='polarity',y='Age',data=df,kind='kde')

<seaborn.axisgrid.JointGrid at 0x7f0230dd6210>



✓ 0s completed at 1:15 AM

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