```
12/15/22, 2:52 PM
                                                                                         covid - Jupyter Notebook
  In [1]:
  import re
  import numpy as np
  import pandas as pd
  import seaborn as sns
  import matplotlib.pyplot as plt
  %matplotlib inline
  from wordcloud import WordCloud, STOPWORDS
  import nltk
  from nltk.corpus import stopwords
  from nltk.tokenize import sent tokenize
  from nltk.tokenize import word_tokenize
  from nltk.stem import WordNetLemmatizer
  lemmatizer = WordNetLemmatizer()
  stopword = set(stopwords.words('english'))
  In [2]:
  import warnings
  warnings.filterwarnings('ignore')
  In [3]:
  df = pd.read_csv('covid_abstracts.csv')
  In [4]:
  df.head()
  Out[4]:
                                                title
                                                                                             abstract
   0
        Real-World Experience with COVID-19 Including...
                                                          This article summarizes the experiences of COV... https://pubmed.ncbi.nlm.nih.gov/35008137
      Successful outcome of pre-engraftment COVID-19...
                                                          Coronavirus disease 2019 COVID-19 caused by... https://pubmed.ncbi.nlm.nih.gov/35008104
       The impact of COVID-19 on oncology professiona... BACKGROUND COVID-19 has had a significant imp... https://pubmed.ncbi.nlm.nih.gov/35007996
   3
          ICU admission and mortality classifiers for CO ...
                                                         The coronavirus disease 2019 COVID-19 which ... https://pubmed.ncbi.nlm.nih.gov/35007991
          Clinical evaluation of nasopharyngeal midturb...
                                                            In the setting of supply chain shortages of na... https://pubmed.ncbi.nlm.nih.gov/35007959
  In [5]:
  df.tail()
  Out[5]:
                                                   title
                                                                                               abstract
                                                                                                                                           url
   9995
           Rooming-in Breastfeeding and Neonatal Follow-...
                                                        INTRODUCTION Due to growing evidence suggesti... https://pubmed.ncbi.nlm.nih.gov/34851815
            Acute Retinal Necrosis from Reactivation of Va...
                                                            PURPOSE To report a case of acute retinal nec... https://pubmed.ncbi.nlm.nih.gov/34851795
   9997
         Acute Abducens Nerve Palsy Following the Secon...
                                                             The authors report the case of an otherwise he... https://pubmed.ncbi.nlm.nih.gov/34851785
   9998
            Planning and Implementing the Protocol for Psy...
                                                             The present study aims to plan the protocol fo... https://pubmed.ncbi.nlm.nih.gov/34851781
   9999
             Prolonged corrected QT interval in hospitalize...
                                                            OBJECTIVE To evaluate the association of a pr... https://pubmed.ncbi.nlm.nih.gov/34851769
  In [6]:
```

```
df.shape
Out[6]:
(10000, 3)
In [7]:
df.columns
Out[7]:
Index(['title', 'abstract', 'url'], dtype='object')
In [8]:
df.duplicated().sum()
```

Out[8]:

0

```
In [9]:
df.isnull().sum()
Out[9]:
title
             a
            a
abstract
url
             0
dtype: int64
In [10]:
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999 \,
Data columns (total 3 columns):
                Non-Null Count Dtype
 # Column
---
     -----
                -----
 0 title
                10000 non-null object
 1 abstract 10000 non-null object
2 url 10000 non-null object
                10000 non-null object
dtypes: object(3)
memory usage: 234.5+ KB
In [11]:
df.nunique()
Out[11]:
             10000
title
abstract
             10000
url
            10000
dtype: int64
In [12]:
def remove_punctuation(text):
    # punctuations except -
punc ='''?!.,:;_-[](){}'"`~|\/@#$%^&+=*'''
    for i in text:
        if i in punc:
             text = text.replace(i, ' ')
    return text.strip()
def word_count(text):
    # word tokenization
    lst = word_tokenize(text)
    return len(lst)
def preprocess(text):
    # Lower casing
    text=text.lower()
    # stopword removal
    text = [word for word in text.split(' ') if word not in stopword]
text=" ".join(text)
    # Lemmatization
    text = [lemmatizer.lemmatize(word) for word in text.split(' ')]
    text = " ".join(text)
    # remove extra spaces
text = re.sub("\s\s+", " ", text)
    return text.strip()
In [13]:
```

```
# apply functions
df['title']=df['title'].apply(remove_punctuation)
df['wc_title']=df['title'].apply(word_count)
```

In [14]:

df

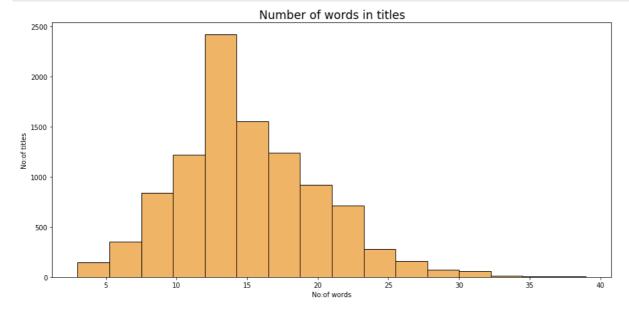
Out[14]:

	title	abstract	url	wc_title
0	Real-World Experience with COVID-19 Including	This article summarizes the experiences of COV	https://pubmed.ncbi.nlm.nih.gov/35008137	21
1	Successful outcome of pre-engraftment COVID-19	Coronavirus disease 2019 COVID-19 caused by	https://pubmed.ncbi.nlm.nih.gov/35008104	16
2	The impact of COVID-19 on oncology professiona	BACKGROUND COVID-19 has had a significant imp	https://pubmed.ncbi.nlm.nih.gov/35007996	19
3	ICU admission and mortality classifiers for CO	The coronavirus disease 2019 COVID-19 which	https://pubmed.ncbi.nlm.nih.gov/35007991	18
4	Clinical evaluation of nasopharyngeal midturb	In the setting of supply chain shortages of na	https://pubmed.ncbi.nlm.nih.gov/35007959	14
9995	Rooming-in Breastfeeding and Neonatal Follow	INTRODUCTION Due to growing evidence suggesti	https://pubmed.ncbi.nlm.nih.gov/34851815	12
9996	Acute Retinal Necrosis from Reactivation of Va	PURPOSE To report a case of acute retinal nec	https://pubmed.ncbi.nlm.nih.gov/34851795	14
9997	Acute Abducens Nerve Palsy Following the Secon	The authors report the case of an otherwise he	https://pubmed.ncbi.nlm.nih.gov/34851785	13
9998	Planning and Implementing the Protocol for Psy	The present study aims to plan the protocol fo	https://pubmed.ncbi.nlm.nih.gov/34851781	17
9999	Prolonged corrected QT interval in hospitalize	OBJECTIVE To evaluate the association of a pr	https://pubmed.ncbi.nlm.nih.gov/34851769	20

10000 rows × 4 columns

In [15]:

```
# plot
plt.figure(figsize=(15,7))
ax=sns.histplot(x='wc_title', data=df, bins=16, color='#eb9b34')
plt.title('Number of words in titles',size='xx-large')
plt.xlabel('No:of words')
plt.ylabel('No:of titles')
plt.show()
```

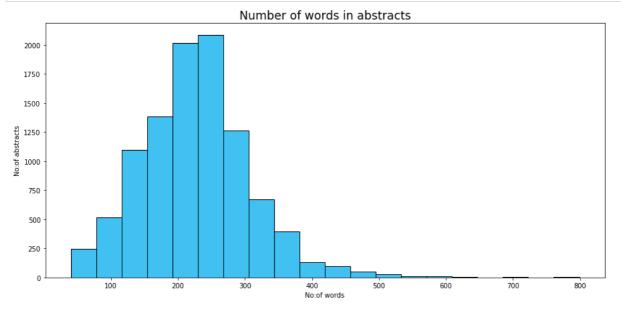


In [16]:

```
# apply functions
df['abstract']=df['abstract'].apply(remove_punctuation)
df['wc_abstract']=df['abstract'].apply(word_count)
```

```
In [17]:
```

```
# plot
plt.figure(figsize=(15,7))
ax=sns.histplot(x='wc_abstract', data=df, bins=20, color='#00acee')
plt.title('Number of words in abstracts',size='xx-large')
plt.xlabel('No:of words')
plt.ylabel('No:of abstracts')
plt.show()
```



In [18]:

```
# preprocess text
df['cleaned_title']=df['title'].apply(preprocess)
```

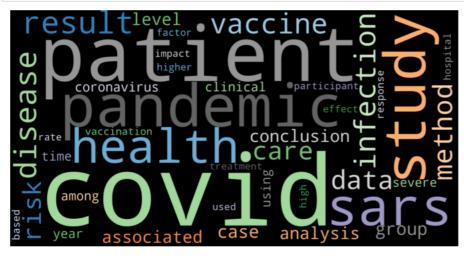
In [19]:

```
Careeffect disease analysis p social healthcare worker Dandem 1 Cypus among variant mental report response using among among infections tudy based variant mental coronavirus based vaccinationsystematic healthcare worker Dandem 1 Cypus among among among among among based vaccinationsystematic healthcare worker Dandem 1 Cypus among among among among among among based vaccinationsystematic healthcare worker Dandem 1 Cypus among among among among among based vaccinationsystematic healthcare worker Dandem 1 Cypus among among among among among based vaccinationsystematic healthcare worker Dandem 1 Cypus among among among among among based vaccinationsystematic healthcare worker Dandem 1 Cypus among among
```

In [20]:

```
# preprocess text
df['cleaned_abstract']=df['abstract'].apply(preprocess)
```

In [21]:



In [22]:

df

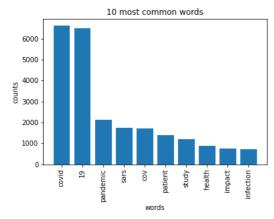
Out[22]:

	title	abstract	url	wc_title	wc_abstract	cleaned_title	cleaned_abstract
0	Real-World Experience with COVID-19 Including	This article summarizes the experiences of COV	https://pubmed.ncbi.nlm.nih.gov/35008137	21	264	real-world experience covid-19 including direc	article summarizes experience covid-19 patient
1	Successful outcome of pre-engraftment COVID-19	Coronavirus disease 2019 COVID-19 caused by	https://pubmed.ncbi.nlm.nih.gov/35008104	16	200	successful outcome pre-engraftment covid-19 hc	coronavirus disease 2019 covid-19 caused sever
2	The impact of COVID- 19 on oncology professiona	BACKGROUND COVID-19 has had a significant imp	https://pubmed.ncbi.nlm.nih.gov/35007996	19	315	impact covid-19 oncology professionals-one yea	background covid-19 significant impact well- be
3	ICU admission and mortality classifiers for CO	The coronavirus disease 2019 COVID-19 which	https://pubmed.ncbi.nlm.nih.gov/35007991	18	299	icu admission mortality classifier covid-19 pa	coronavirus disease 2019 covid-19 caused sever
4	Clinical evaluation of nasopharyngeal midturb	In the setting of supply chain shortages of na	https://pubmed.ncbi.nlm.nih.gov/35007959	14	164	clinical evaluation nasopharyngeal midturbinat	setting supply chain shortage nasopharyngeal n
9995	Rooming-in Breastfeeding and Neonatal Follow	INTRODUCTION Due to growing evidence suggesti	https://pubmed.ncbi.nlm.nih.gov/34851815	12	305	rooming-in breastfeeding neonatal follow-up in	introduction due growing evidence suggesting c
9996	Acute Retinal Necrosis from Reactivation of Va	PURPOSE To report a case of acute retinal nec	https://pubmed.ncbi.nlm.nih.gov/34851795	14	129	acute retinal necrosis reactivation varicella	purpose report case acute retinal necrosis arn
9997	Acute Abducens Nerve Palsy Following the Secon	The authors report the case of an otherwise he	https://pubmed.ncbi.nlm.nih.gov/34851785	13	105	acute abducens nerve palsy following second do	author report case otherwise healthy 65- year-o
9998	Planning and Implementing the Protocol for Psy	The present study aims to plan the protocol fo	https://pubmed.ncbi.nlm.nih.gov/34851781	17	196	planning implementing protocol psychosocial in	present study aim plan protocol providing psyc
9999	Prolonged corrected QT interval in hospitalize	OBJECTIVE To evaluate the association of a pr	https://pubmed.ncbi.nlm.nih.gov/34851769	20	199	prolonged corrected qt interval hospitalized p	objective evaluate association prolonged corre

10000 rows × 7 columns

In [28]:

```
\textbf{from} \ \textbf{sklearn.feature\_extraction.text} \ \textbf{import} \ \textbf{CountVectorizer}
# Helper function
def plot_10_most_common_words(count_data, count_vectorizer):
    import matplotlib.pyplot as plt
    words = count_vectorizer.get_feature_names()
    total_counts = np.zeros(len(words))
    for t in count_data:
        total_counts+=t.toarray()[0]
    count_dict = (zip(words, total_counts))
    count_dict = sorted(count_dict, key=lambda x:x[1], reverse=True)[0:10]
    words = [w[0] for w in count_dict]
    counts = [w[1] for w in count_dict]
    x_pos = np.arange(len(words))
    plt.bar(x_pos, counts,align='center')
    plt.xticks(x_pos, words, rotation=90)
    plt.xlabel('words')
    plt.ylabel('counts')
    plt.title('10 most common words')
    plt.show()
# Initialise the count vectorizer with the English stop words
count_vectorizer = CountVectorizer(stop_words ='english')
# Fit and transform the processed titles
count_data = count_vectorizer.fit_transform(df['cleaned_title'])
# Visualise the 10 most common words
plot_10_most_common_words(count_data, count_vectorizer)
```

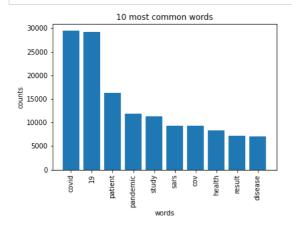


In [29]:

```
count_data_abs = count_vectorizer.fit_transform(df['cleaned_abstract'])
```

In [30]:

plot_10_most_common_words(count_data_abs, count_vectorizer)



In [31]:

```
from sklearn.decomposition import LatentDirichletAllocation as LDA
# Helper function
def print_topics(model, count_vectorizer, n_top_words):
    words = count_vectorizer.get_feature_names()
    for topic_idx, topic in enumerate(model.components_):
    print("\nTopic #%d:" % topic_idx)
    print(" ".join([words[i]
                         for i in topic.argsort()[:-n_top_words - 1:-1]]))
# Tweak the two parameters below (use int values below 15)
number\_topics = 10
number\_words = 10
# Create and fit the LDA model
lda = LDA(n_components=number_topics)
lda.fit(count_data)
# Print the topics found by the LDA model
print("Topics found via LDA:")
print_topics(lda, count_vectorizer, number_words)
Topics found via LDA:
Topic #0:
cl2 496 criticized blackfirst csrc 3x 6172 crhr bedding alzheimer
Topic #1:
cl2 496 ante 14 catarrhal 4750 1976099 chewing commencing 6654
4m 00244 brackish compressible 515 compressibility climax 27186 4750 androgenic
Tonic #3:
4m 00244 british chloroquine compressible 13108567 brackish 3373 bona congregation
4m 00244 brackish 2793 chaudhry clause clip 439 canceled cvds
Topic #5:
2972 corresponded crhr centrifugation analyzing 00244 4m anosmic ante chewing
crhr 00244 4m crgns compressible 0064 battery 0068 albania cvi
4m 00244 british 7274 brackish confuse anthropogenic 2793 ante asw
4m 00244 brackish albacete 27186 autoptic androgenic 30b compressible albania
4m 00244 574 comparedimmune artisan cq brackish creator 1976099 addressing
```