## **#CLASSIFICATION: TRUE VS FALSE AVEC ENTITES\_NOMMEES**

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```
import sys
import pandas as pd
import numpy as np
import sklearn
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import accuracy score, confusion matrix,
classification report, precision recall fscore support as score
from sklearn.model selection import train test split, KFold,
cross val score, GridSearchCV
from sklearn.linear model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive bayes import GaussianNB
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.feature extraction.text import TfidfVectorizer
import nltk
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer, WordNetLemmatizer
from nltk import word tokenize
import re
import pickle
import string
import seaborn as sns
import matplotlib.pyplot as plt
from numpy import vstack
from torch.utils.data import Dataset, DataLoader
from torch.utils.data import random split
from torch import Tensor
from torch.nn import Linear, ReLU, Sigmoid, Module, BCELoss
from torch.optim import SGD
from torch.nn.init import kaiming uniform , xavier uniform
kn= KNeighborsClassifier()
decTree=DecisionTreeClassifier()
gNB=GaussianNB()
svc=SVC()
rf=RandomForestClassifier()
import warnings
warnings.filterwarnings('ignore')
autorisation
```

```
drive.mount('/content/drive')
Drive already mounted at /content/drive; to attempt to forcibly
remount, call drive.mount("/content/drive", force remount=True).
my_local_drive='/content/gdrive/My Drive/Colab Notebooks'
# Ajout du path pour les librairies, fonctions et données
sys.path.append(my_local_drive)
# Se positionner sur le répertoire associé
%cd $my local drive
%ls
%pwd
/content/gdrive/My Drive/Colab Notebooks
'Copie de bonFakeNEWS.ipynb'
'Copie de Copie de Projet ML FakeNEWS TRUE FALSE TEXT+TITRE.ipynb'
'Copie de ml entiteesNommeesTest (1).ipynb'
'Copie de ml entiteesNommeesTest.ipynb'
'Copie de Projet ML FakeNEWS (1).ipynb'
'Copie de Projet ML FakeNEWS.ipynb'
'Copie de Projet ML FakeNEWS TRUE FALSE TEXT+TITRE.ipynb'
'Copie de text title en mm temps.ipynb'
'Copie de version2ml entiteesNommeesTest.ipynb'
'Copie de Vrai Faux.ipynb'
'copie essai title text sur bachir.ipynb'
'essai title text.ipynb'
'Projet ML FakeNEWS TRUE FALSE TEXT AVEC ENTITES.ipynb'
'Projet ML FakeNEWS TRUE FALSE TITRE AVEC ENTITES.ipynb'
'Traitement sémantique'/
Vrai_Faux_avec_entites_nommées.ipynb
{"type": "string"}
La fonction qui sera utilisée pour les prétraitements: MyCleanText
     Mettre le texte en minuscule
     Se débarasser des stopwords
     Se débarasser des nombres
     Stemmatisation
     Lemmatisation ..
La fonction MyshowAllScores prend le y_test et le y_predict, affiche l'accuracy et le
classification report avec la matrice de confusion.
```

# mettre en minuscule #enlever les stopwords #se debarasser des nombres

```
#stemmatisation
#lemmatisation
nltk.download('wordnet')
nltk.download('stopwords')
nltk.download('punkt')
#liste des stopwords en anglais
stop words = set(stopwords.words('english'))
def MyCleanText(X,
                lowercase=False, #mettre en minuscule
                removestopwords=False, #supprimer les stopwords
                removedigit=False, #supprimer les nombres
                getstemmer=False, #conserver la racine des termes
                getlemmatisation=False #lemmatisation des termes
                ):
  #conversion du texte d'entrée en chaîne de caractères
    sentence=str(X)
    #suppression des caractères spéciaux
    sentence = re.sub(r'[^\w\s]',' ', sentence)
    # suppression de tous les caractères uniques
    sentence = re.sub(r'\s+[a-zA-Z]\s+', ' ', sentence)
    # substitution des espaces multiples par un seul espace
    sentence = re.sub(r'\s+', ' ', sentence, flags=re.I)
    # decoupage en mots
    tokens = word tokenize(sentence)
    if lowercase:
          tokens = [token.lower() for token in tokens]
    # suppression ponctuation
    table = str.maketrans('', '', string.punctuation)
    words = [token.translate(table) for token in tokens]
    # suppression des tokens non alphabetique ou numerique
    words = [word for word in words if word.isalnum()]
    # suppression des tokens numerique
    if removedigit:
        words = [word for word in words if not word.isdigit()]
    # suppression des stopwords
    if removestopwords:
        words = [word for word in words if not word in stop words]
```

```
# lemmatisation
   if getlemmatisation:
      lemmatizer=WordNetLemmatizer()
      words = [lemmatizer.lemmatize(word)for word in words]
   # racinisation
   if getstemmer:
      ps = PorterStemmer()
      words=[ps.stem(word) for word in words]
   sentence= ' '.join(words)
   return sentence
def MyshowAllScores(y_test,y_pred):
 classes= np.unique(y test)
 print("Accuracy : %0.3f"%(accuracy score(y test,y pred)))
 print("Classification Report")
 print(classification report(y test,y pred,digits=5))
 cnf matrix = confusion_matrix(y_test,y_pred)
 disp=ConfusionMatrixDisplay(cnf matrix, display labels=classes)
 disp.plot()
[nltk data] Downloading package wordnet to /root/nltk data...
           Package wordnet is already up-to-date!
[nltk data]
[nltk data] Downloading package stopwords to /root/nltk_data...
           Package stopwords is already up-to-date!
[nltk data]
[nltk data] Downloading package punkt to /root/nltk data...
           Package punkt is already up-to-date!
[nltk data]
    La classe TextNormalizer qui contiendra la fonction MyCleanText.
    Fit_transform de mon corpus propre.
#.....Etape 1 :
prétraitement du
#fit transform de mon corpus propre
from sklearn.base import BaseEstimator, TransformerMixin
```

```
class TextNormalizer(BaseEstimator, TransformerMixin):
    def init (self,
                 removestopwords=False, # suppression des stopwords
                 lowercase=False,# passage en minuscule
                 removedigit=False, # supprimer les nombres
                 getstemmer=False,# racinisation des termes
                 getlemmatisation=False # lemmatisation des termes
                ):
        self.lowercase=lowercase
        self.getstemmer=getstemmer
        self.removestopwords=removestopwords
        self.getlemmatisation=getlemmatisation
        self.removedigit=removedigit
    def transform(self, X, **transform params):
        # Nettovage du texte
        X=X.copy() # pour conserver le fichier d'origine
        return [MyCleanText(text,lowercase=self.lowercase,
                            getstemmer=self.getstemmer,
                            removestopwords=self.removestopwords,
                            getlemmatisation=self.getlemmatisation,
                            removedigit=self.removedigit) for text in
X1
    def fit(self, X, y=None, **fit params):
        return self
    def fit transform(self, X, y=None, **fit params):
        return self.fit(X).transform(X)
    def get params(self, deep=True):
        return {
            'lowercase':self.lowercase,
            'getstemmer':self.getstemmer,
            'removestopwords':self.removestopwords,
            'getlemmatisation':self.getlemmatisation,
            'removedigit':self.removedigit
        }
    def set params (self, **parameters):
        for parameter, value in parameters.items():
            setattr(self,parameter,value)
        return self
```

## ##Etape 1 : Préparer les données

- Load et preparer les données à partir des 2 fichiers csv
- Sélectionner que les lignes où on a True, False

```
#Ici je cherche à séléctionner que les labels TRUE et FALSE, donc les
LIGNES qui contiennent au rating TRUE et FALSE uniquement, le reste on
enlève
dftrain = pd.read csv("/content/gdrive/MyDrive/newsTrain2.csv",
names=['id','text','title','rating'], header=0,sep=',',
encoding='utf8')
dftrain.reset index(drop = True, inplace = True)
dftrain2 = pd.read csv("/content/gdrive/MyDrive/newsTrain -
_newsTrain.csv", names=['id','text','title','rating'],
header=0,sep=',', encoding='utf8')
dftrain2.reset index(drop = True, inplace = True)
# concaténer les deux dataframes en ajoutant les lignes du deuxième à
la fin du premier
dftrain = pd.concat([dftrain, dftrain2], ignore index=True)
dftrain = dftrain.loc[dftrain['rating'].isin(['TRUE', 'FALSE'])]
#dftest=pd.read csv("/content/gdrive/MyDrive/newsTest.csv",names=['id'
,'text','title','rating','veracity'], header=0,sep=',',
encoding='utf8')
print("Echantillon de mon dataset \n")
print(dftrain.sample(n=10))
print("\n")
print("Quelques informations importantes \n")
dftrain.info()
#print(dftest.head())
X text=dftrain.iloc[0:,1:2]
print("le type de X_test est" ,X_text.columns)
X title=dftrain.iloc[0:,2:3]
print("le texte est")
display(X text)
print("le titre est")
display(X title)
#X test=dftest.iloc[1:, :4]
y=dftrain.iloc[0:,-1]
print("voici la dernière case")
display(v)
#y test=dftest.iloc[1:, -1]
#y = y train.ravel()
print("la taille de X text est",X text.shape)
print("la taille de y_train est " ,y.shape)
print("les valeurs de TRUE et FALSE sont " ,y.value_counts())
Echantillon de mon dataset
```

```
id
                                                              text \
697
      ab66dbab
                Bacon, burgers and sausages are as big a cance...
                Commander-in-Chief of the Defence Services Min...
644
      45d1505d
2406
      e72c4292
                Lord Neuberger said that judicial policy on ve...
2211
     a1334b14
                War-torn eastern regions of Ukraine have no la...
865
      e2bc4450
                0 DECATUR, Alabama — Police were recently tipp...
1921
     3971d0e8
                HOUSTON - In the weeks since President Obama's...
2446
     acd7cfb8
                Huge reductions in meat-eating are essential t...
2381
     7212bc8a
                The Labour Party has accused the Government of...
11
      2ec747ae
                Submitted by Great Game India Last year a mys...
1202
      0a68da63
                Australia's Great Barrier Reef has worst coral...
                                                  title rating
697
     Trump aims new blasts at McCain, claims credit...
                                                         FALSE
644
                   Myanmar announces state of emergency
                                                         FALSE
2406
     Veils should not be allowed in court, says Bri...
                                                          TRUE
     Look No Further, The Best Doctor Strange in th...
2211
                                                         FALSE
      Georgia to Review Mail-in Ballot Signatures to...
865
                                                         FALSE
1921
     Ex-ambassador McFaul switches to Russian to ce...
                                                         FALSE
2446
     Huge reduction in meat-eating 'essential' to a...
                                                          TRUE
2381
      Chancellor failing patients with his £30 billi...
                                                         FALSE
      Did China Steal Coronavirus From Canada And We...
11
                                                         FALSE
1202
     Australia's Great Barrier Reef has worst coral...
                                                          TRUE
Quelques informations importantes
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1578 entries, 0 to 2527
Data columns (total 4 columns):
#
     Column Non-Null Count
                             Dtype
     -----
             _____
             1578 non-null
 0
     id
                             object
 1
             1578 non-null
                             object
     text
 2
     title
             1554 non-null
                             object
 3
     rating 1578 non-null
                             object
dtypes: object(4)
memory usage: 61.6+ KB
le type de X test est Index(['text'], dtype='object')
le texte est
0
      Distracted driving causes more deaths in Canad...
3
      But things took a turn for the worse when riot...
4
      It's no secret that Epstein and Schiff share a...
6
      November 23, 2019 The U.S. Food and Drug Admi...
7
     Trump confirms this was a bombing, not an acci...
2523
     More than four million calls to the taxman are...
2524
     More under-18s are being taken to court for se...
```

```
The Government's much vaunted Help to Buy Isa ...
2525
      The late Robin Williams once called cocaine "G...
2526
     The late Robin Williams once called cocaine "G...
2527
[1578 rows x 1 columns]
le titre est
                                                   title
      You Can Be Fined $1,500 If Your Passenger Is U...
0
      Obama's Daughters Caught on Camera Burning US ...
3
4
      Leaked Visitor Logs Reveal Schiff's 78 Visits ...
      FDA Shocking Study: Cells Used In Vaccines Con...
6
7
      Israel Hits Beirut with Nuclear Missile, Trump...
. . .
2523
      Taxman fails to answer four million calls a ye...
      Police catch 11-year-olds being used to sell d...
2524
      Help to Buy Isa scandal: 500,000 first-time bu...
2525
2526
               A coke-snorting generation of hypocrites
2527
               A coke-snorting generation of hypocrites
[1578 rows x 1 columns]
voici la dernière case
0
        FALSE
3
        FALSE
        FALSE
4
6
        FALSE
7
        FALSE
        . . .
2523
         TRUE
2524
         TRUE
2525
        FALSE
2526
         TRUE
2527
         TRUE
Name: rating, Length: 1578, dtype: object
la taille de X text est (1578, 1)
la taille de y train est (1578,)
les valeurs de TRUE et FALSE sont FALSE
                                             1156
TRUE
          422
Name: rating, dtype: int64
```

Le jeu de données étant déséquilibré, on a pensé à appliquer le downsampling pour équilibrer nos données. On fait en sorte que le nombre de lignes avec en sortie FAUX soit égal au nombre de lignes avec en sortie true.

#On applique du sous-échantillonnage (downsampling) : car on a plus de FALSE (578) que des TRUE (211)

```
# Séparer les classes en deux dataframes
df_false = dftrain[dftrain['rating'] == 'FALSE']
df true = dftrain [dftrain['rating'] == 'TRUE']
# Sous-échantillonner la classe majoritaire (FALSE) pour obtenir un
nombre égal d'échantillons pour chaque classe
df false subsampled = df false.sample(n=len(df true), random state=42)
# Concaténer les deux dataframes
dftrain = pd.concat([df false subsampled, df true])
# Mélanger aléatoirement les données
dftrain = dftrain.sample(frac=1, random state=42)
X_text=dftrain["text"]
X title=dftrain["title"]
y=dftrain.iloc[0:,-1]
print("la taille de X text est",X text.shape)
print("\n")
print("la taille de X title est",X title.shape)
print("\n")
print("la taille de y train est " ,y.shape)
print("\n")
print("les valeurs de TRUE et FALSE maintenant sont
  ,y.value counts())
la taille de X text est (844,)
la taille de X title est (844,)
la taille de y train est (844,)
les valeurs de TRUE et FALSE maintenant sont TRUE
                                                       422
FALSE
         422
Name: rating, dtype: int64
```

## ##Etape 2 : Classification selon la colonne TEXT + extraction/application d'entités nommées

Tests de Classification pour trouver le meilleur classifieur, les meilleurs paramètres de prétraitement et les meilleurs hyper\_paramètres.

Avant de classifier on applique les entités nommées, Cette fonction utilise Spacy pour extraire les entités nommées à partir d'une liste de textes et les stocke dans une liste pour être affichées avec leurs labels correspondants.

```
nlp = spacy.load("en core web sm")
texte=dftrain['text']
#Créer une liste pour stocker les entités nommées
entities = []
# Traiter chaque texte individuellement et ajouter les entités à la
liste
for phrase in texte:
    phrase str = str(phrase) # Convertir l'élément en chaîne de
caractères
    doc = nlp(phrase str)
    for ent in doc.ents:
        entities.append((ent.text, ent.label ))
# Afficher les entités nommées
for entity in entities:
    print(entity[0], entity[1])
Le flux de sortie a été tronqué et ne contient que les 5000 dernières
lignes.
EPA ORG
UK GPE
10 April 2021 DATE
The Death Gun Salute ORG
the Honourable Artillery Company ORG
Britain GPE
Prince Philip GPE
the The Tower of London AFP FAC
Getty UK ORG
9 April 2021 DATE
Buckingham Palace FAC
London GPE
Britain GPE
Prince Philip GPE
Elizabeth II's PERSON
Prince Philip PERSON
more than a month DATE
April 9 DATE
Buckingham Palace FAC
99 DATE
Getty Images ORG
UK GPE
8 April 2021 DATE
James King PERSON
4:05 TIME
REUTERS ORG
```

UK GPE

7 April 2021 DATE

the Independent Workers' Union of Great Britain ORG

Deliveroo GPE

London GPE

UK GPE

6 April 2021 DATE

Seaham Lighthouse ORG

Durham PA GPE

UK GPE

5 April 2021 DATE

Lusamba Katalay ORG

Belly Mujinga GPE

Victoria GPE

London GPE

first ORDINAL

Belly Mujinga GPE

London GPE

Victoria GPE

UK GPE

4 April 2021 DATE

Easter Sunday at EVENT

Hengistbury Head FAC

UK GPE

3 April 2021 DATE

Bill PERSON

London GPE

EPA ORG

UK GPE

2 April 2021 DATE

Bamburgh Croquet ORG

COVID-19 ORG

Northumberland GPE

Britain GPE

UK GPE

1 April 2021 DATE

St Nicholas' Park PERSON

Warwick GPE

UK GPE

this week DATE

Easter ORG

PA UK ORG

31 March 2021 DATE

the National Covid Memorial Wall ORG

London GPE

EPA ORG

UK GPE

30 March DATE

Jamie Klingler PERSON

Reclaim These Streets ORG

Clapham Common FAC

London GPE

the Chief Inspector of Constabulary ORG

Thomas Winsor PERSON

Metropolitan Police ORG

PA UK ORG

29 March 2021 DATE

the Mersey Bowman Lawn Tennis Club ORG

Liverpool GPE

England GPE

March 29, 2021 DATE

England GPE

third ORDINAL

up to six CARDINAL

England GPE

Monday DATE

second ORDINAL

AFP ORG

Getty Images ORG

UK GPE

28 March 2021 DATE

Tenby Lifeboat ORG

Tenby GPE

Wales Reuters ORG

UK GPE

27 March 2021 DATE

Manchester Reuters ORG

UK GPE

26 March 2021 DATE

Andy Warhol PERSON

Elizabeth II PERSON

Reigning Queens ORG

1985 DATE

Elizabeth II PERSON

1985 DATE

Christie's ORG

London GPE

UK GPE

25 March 2021 DATE

Charlton Lido PERSON

London GPE

PA UK ORG

24 March 2021 DATE

Scottish Green Party ORG

Patrick Harvie PERSON

Lorna Slater PERSON

Edinburgh GPE

PA GPE

UK GPE

23 March 2021 DATE

Westbourne House ORG

Sheffield GPE

the National Reflection day EVENT

first ORDINAL

first ORDINAL

UK GPE

Tom Maddick PERSON

UK GPE

22 March 2021 DATE

Couple CARDINAL

Sean Henry PERSON

Newbiggin GPE

the-Sea LOC

Northumberland GPE

UK GPE

21 March 2021 DATE

Jersey GPE

meteor ORG

UK GPE

20 March 2021 DATE

London GPE

UK GPE

19 March 2021 DATE

St. James's Park GPE

London GPE

Britain GPE

UK GPE

18 March 2021 DATE

Highland PERSON

the National Trust's ORG

Cambridgeshire GPE

UK GPE

17 March 2021 DATE

early in the morning TIME

PA GPE

UK GPE

16 March 2021 DATE

60-metre OUANTITY

12-year-old DATE

Ansha PERSON

Ethiopia GPE

WaterAid ORG

Whitby Beach GPE

Yorkshire GPE

hours TIME

each day DATE

UK GPE

15 March 2021 DATE

2021 DATE

Teresa Solar ORG

Exchange Flags ORG one CARDINAL 11th ORDINAL Liverpool Biennial ORG March 20, 2021 DATE nine CARDINAL PA Photo ORG Monday March 15, 2021 DATE PA Photo ORG second ORDINAL UK GPE 14 March 2021 DATE Sarah Everard PERSON Clapham Common FAC Sarah Everard's PERSON Parliament Square FAC London AFP PERSON Getty UK ORG 13 March 2021 DATE Sarah Everard PERSON Clapham Common FAC London GPE Getty Images ORG UK GPE 12 March DATE 2021 DATE Angela Rayner PERSON the Labour Party ORG Jonathan Ashworth PERSON Shadow Health WORK\_OF\_ART London GPE 1 percent PERCENT Thursday DATE May PA DATE UK GPE 11 March 2021 DATE Newhaven GPE England AFP/ PERSON UK GPE 10 March 2021 DATE Great Chart Golf FAC Leisure in Ashford ORG UK GPE 9 March 2021 DATE Kielder Water FAC Northumberland GPE the Border Mires ORG 50 year old DATE the Border Mires in WORK OF ART

Northumberland GPE

UK GPE

8 March 2021 DATE

the Jewellery Quarter Academy ORG

Birmingham GPE

the West Midlands LOC

England GPE

first ORDINAL

two months DATE

first ORDINAL

UK GPE

7 March 2021 DATE

NHS ORG

NHS Workers ORG

Unite ORG

Downing Street FAC

London GPE

1% PERCENT

UK GPE

6 March 2021 DATE

Scottish NORP

Ibrox Stadium FAC

UK GPE

March 2021 DATE

Sutton Manor GPE

St Helens PERSON

a clear DATE

Spring day DATE

PA GPE

UK GPE

4 March 2021 DATE

Health ORG

Matt Hancock PERSON

the Centre for Virus Research ORG

Glasgow PA GPE

UK GPE

March 2021 DATE

Exchequer ORG

Rishi Sunak ORG

11 Downing Street ORG

London GPE

the House of Commons ORG

Budget PA UK ORG

2 March 2021 DATE

the Carbis Bay Estate PRODUCT

G7 CARDINAL

Getty UK ORG

1 March 2021 DATE

A J Thompson & Sons ORG

the Romney Marsh ORG

Kent PERSON

the first day DATE

UK GPE

28 CARDINAL

February 2021 DATE

Scarborough GPE

North Yorkshire GPE

UK GPE

27 February 2021 DATE

Tom Moore PERSON

the Armed Forces at Bedford Crematorium ORG

Bedford GPE

EPA ORG

UK GPE

26 February 2021 DATE

New Brighton GPE

Wirral PA UK ORG

25 February 2021 DATE

Shadow PRODUCT

Anneliese NORP

the Institute of Global Prosperity ORG

Labour Party ORG

London GPE

UK GPE

24 February 2021 DATE

Britain GPE

Health ORG

Matt Hancock PERSON

Downing Street FAC

London GPE

Britain GPE

UK GPE

23 CARDINAL

February 2021 DATE

Two CARDINAL

Boscombe GPE

Dorset PA GPE

Dean Haydon PERSON

national counterterror policing ORG

CT Policing ORG

London GPE

11 CARDINAL

Theresa May PERSON

Downing St PERSON

last year's DATE

more than 650 CARDINAL

about a third CARDINAL

the start of 2017 DATE

Haydon PERSON

the last year DATE

the Counter Terrorism and Border Security Bill ORG

two CARDINAL

Terrorism Acts ORG

0800 CARDINAL

321 CARDINAL

Taylor Zundel PERSON

Salt Lake City GPE

at least one CARDINAL

82-Year-Old DATE

30 Years DATE

only two CARDINAL

at least one CARDINAL

two CARDINAL

Democratic Party ORG

Entire Trays of Mail ORG

Wisconsin GPE

Hunter Ribarchak PERSON

7 DATE

the Iron Workers of America's ORG

Labor Day Parade EVENT

Pittsburgh GPE

Pa. GPE

Monday, Sept. 1, 2008 DATE

Labor Day DATE

this year DATE

American NORP

25 million CARDINAL

American NORP

the worst decade DATE

American NORP

a century DATE

Commerce Department ORG

the last decade DATE

the decade of the Great Depression DATE

4 percent PERCENT

the last ten years DATE

5 percent PERCENT

1929 to 1939 DATE

Big American NORP

the United States GPE

American NORP

the United States GPE

the Supreme Court's ORG

American NORP

350 CARDINAL

American NORP

11 percent PERCENT

last year DATE

\$9.3 million MONEY

The Wall Street Journal ORG

Hay Group ORG

19.7 percent PERCENT

2008 DATE

2009 DATE

March, 2009 DATE

Ford ORG

Alan Mulally PERSON

an estimated \$16 million MONEY

Ford ORG

UAW ORG

Ford ORG

roughly half CARDINAL

Ford ORG

Mulally's ORG

2009 DATE

over \$200 million MONEY

the Great Depression EVENT

the Great Depression EVENT

American NORP

70 percent PERCENT

Labor Day EVENT

Washington GPE

Republicans NORP

Republican NORP

Indiana GPE

Ohio GPE

Maine GPE

Wisconsin GPE

this Labor Day DATE

American NORP

Labor Day EVENT

1979-2000 CARDINAL

late twentieth century DATE

twentieth century DATE

2 CARDINAL

the twenty-first century DATE

Unreliable Circumstantial ORG

Arctic sea LOC

the twentieth century DATE

overthe past 100 years DATE

England GPE

up to £ MONEY

120 MONEY

28 days DATE

just over half CARDINAL

David Cameron PERSON

BBC Breakfast ORG

the Conservative Party ORG

60 MONEY

England GPE

21 days DATE

28 days DATE

David Cameron PERSON

BBC Breakfast ORG

the Conservative Party ORG

Manchester GPE

The Press Association ORG

earlier this year DATE

16,430 CARDINAL

England GPE

last year DATE

About three-quarters - 12,479 CARDINAL

9,214 CARDINAL

172 MONEY

Cameron PERSON

first ORDINAL

September 2011 DATE

April 2012 DATE

Chris Keates PERSON

the teachers union ORG

NASUWT PERSON

England GPE

Schools PERSON

Update ORG

8 October 2015 DATE

28 days DATE

Francis PERSON

this morning TIME

Conservative Beaver ORG

Pope Francis PERSON

TweetDeck ORG

Francis PERSON

Federal Custody ORG

Italy GPE

FBI ORG

10 minutes ago TIME

Vatican FAC

January 10, 2021 DATE

VATICAN ORG

Francis PERSON

Jorge Mario Bergoglio PERSON

Saturday DATE

Vatican FAC

Tim Young PERSON

January 10, 2021 DATE

The Italian National Prosecutor's Office ORG

Italian NORP

Vatican FAC

Francis PERSON

Federal Agents ORG

Italy GPE

Interpol ORG

FBI ORG

Interpol ORG

Giuseppe Governale PERSON

Italy GPE

Francis PERSON

Vatican FAC

Vatican GPE

Italy GPE

Europe LOC

Italy GPE

Giuseppe Governale PERSON

Vatican FAC

Vatican FAC

# CARDINAL

Pakistan GPE

Vatican FAC

@anoncatnews ORG

January 10, 2021 DATE

Vatican FAC

January 10, 2021 DATE

first ORDINAL

Interpol ORG

first ORDINAL

first ORDINAL

@DennisseLisseth January 10, 2021 DATE

EU ORG

Canada GPE

US GPE

Italian NORP

Google ORG

first ORDINAL

Buckingham Palace FAC

David Cameron PERSON

David PERSON

David PERSON

David Cameron PERSON

one CARDINAL

Conservative NORP

Unionist Party ORG

England GPE

Scotland GPE

Wales GPE

Northern Ireland GPE

the United Kingdom GPE

9 years earlier DATE

Britain GPE

Britain GPE

Westminster GPE

Great Britain GPE

the European Union ORG

Britain GPE

Britain GPE

163 CARDINAL

George Soros For PERSON

thousands CARDINAL

Scottish NORP

Criminal Record FAC

Invalid ORG

2015 DATE

a few months DATE

56 CARDINAL

June 8 DATE

Sturgeon ORG

second ORDINAL

Scotland GPE

Sturgeon PERSON

SNP ORG

Scotland GPE

SNP ORG

NHS ORG

Scotland GPE

1billion MONEY

the SNP Government ORG

Scotland GPE

Sturgeon ORG

Scotland GPE

more than 2500 CARDINAL

David Cheskin/PA Wire PERSON

8 CARDINAL

more than six months DATE

474 per cent MONEY

SNP ORG

almost 400 CARDINAL

Scotland GPE

Labour ORG

England GPE

Scotland GPE

SNP ORG

4000 CARDINAL

1000 CARDINAL

2007 DATE

SNP ORG

Labour ORG

England GPE

Scotland GPE

SNP ORG

Scotland GPE

one CARDINAL

Scotland GPE

Scotland GPE

the past year DATE

SNP ORG

1000 CARDINAL

Labour ORG

Scotland GPE

ScotRail ORG

14-year DATE

Labour ORG

SNP ORG

ScotRail ORG

Dutch NORP

Abellio PERSON

Labour ORG

the age of 60 DATE

SNP ORG

Sturgeon ORG

today DATE

second ORDINAL

Labour ORG

June 8 DATE

The Novel Coronavirus Infecting ORG

China GPE

Army ORG

Coronavirus LOC

China GPE

BIO-WEAPON ORG

the Wuhan Virology Laboratory ORG

China GPE

People's Liberation Army ORG

Nanjiang Command ORG

Bat PERSON

the Chinese Army ORG

Two CARDINAL

HIV-1 CARDINAL

Bat PERSON

ACES2 PERSON

BIO-WEAPON PROOF ORG

Bat PERSON

the National Center for Biotechnology Information ORG

the US National Library of Medicine ORG

the year 2018 DATE

China GPE

Institute of Military Medicine ORG

Nanjing Command ORG

the National Center for Biotechnology Information ORG

Bat PERSON

AVP 70833.1 CARDINAL

the National Center for Biotechnology Information ORG

the U.S. Library of Medicine FAC

Coronavirus EVENT

National Center for Biotechnology Information ORG

January of this year DATE

the Shanghai Public Heath Clinical Center ORG

BLAST Basic Local Alignment Search Tool ORG

the U.S. National Library of Medicine ORG

two CARDINAL

100% PERCENT

2019 DATE

2019 - nCoV. QUANTITY

2019 DATE

nCoV ORG

4 CARDINAL

2019 DATE

4 CARDINAL

2019 DATE

nCoV ORG

4 CARDINAL

2019 DATE

HIV-1 PRODUCT

Bat PERSON

the Virus Envelope ORG

100% PERCENT

2018 DATE

Bat PERSON

Coronavirus PERSON

China GPE

People's Liberation Army ORG

the Institute of Military Medicine ORG

Nanjiang Command ORG

the year 2018 DATE

two years later DATE

as day DATE

China GPE

Bat PERSON

LOSE ORG

China GPE

the first month DATE

China GPE

QUARANTINED PERSON

more than fifty-six MILLION CARDINAL

56,000,000 CARDINAL

nineteen 19 DATE

China GPE

only a few hundred CARDINAL

only about 25 CARDINAL

China GPE

19 CARDINAL

56 Million CARDINAL

China GPE

China GPE

China GPE

China GPE

FACTS ORG

eighty-three percent PERCENT

83% PERCENT

100 CARDINAL

EXPOSED ORG

83 CARDINAL

AIR ORG

five days DATE

twenty eight days DATE

a couple minutes, hours TIME

A few minutes later TIME

One CARDINAL

over 100 years DATE

the Spanish Flu of 1918 EVENT

Spanish NORP

Spanish NORP

1918-1919 DATE

the United States GPE

first ORDINAL

spring 1918 DATE

about 500 million CARDINAL

one-third CARDINAL

at least 50 million CARDINAL

about 675,000 CARDINAL

the United States GPE

one CARDINAL

1918-1919 DATE

Spanish LANGUAGE

1.4 CARDINAL

2.8 CARDINAL

2 CARDINAL

1.4 CARDINAL

2.8 CARDINAL

Bio-Weapon ORG

R4.08 PRODUCT

FAAAAAAR Worse ORG

1918 DATE

Spanish Flu EVENT

Satistia ORG

U.S. GPE

1975-2017 DATE

1975 DATE

about 1.5 million CARDINAL

U.S. GPE

2017 DATE

just about 931 thousand CARDINAL

9, 2019 DATE

TrendingEconomics, Hospital ORG

1,000 CARDINAL

United States GPE

2.9 CARDINAL

2013 DATE

World Bank ORG

the Infection Rate ORG

83% PERCENT

1,000 CARDINAL

this Bio-Weapon ORG

830 CARDINAL

2.9 CARDINAL

US GPE

830 CARDINAL

2.9 CARDINAL

2.9 CARDINAL

65-70% PERCENT

56 Million CARDINAL

19 CARDINAL

Quarantined NORP

days DATE

Coronavirus LOC

Quarantined NORP

China GPE

Filter Masks: People WORK\_OF\_ART

Mexico GPE

Hemorrhagic ORG

China GPE

Nailed-in PERSON

One CARDINAL

Hebei GPE

five days DATE

Shaanxi GPE

China GPE

China GPE

only about 30 Days DATE

just 30 days DATE

ONE CARDINAL

ONE CARDINAL

China GPE

WILL GPE

America GPE

Today DATE

February 1 DATE

California GPE

seventh ORDINAL

the United States GPE

Santa Clara County GPE

the Centers for Disease Control and Prevention ORG

Friday DATE

afternoon TIME

two CARDINAL

California GPE

two CARDINAL

Illinois GPE

Arizona GPE

one CARDINAL

Washington GPE

4:43 PM EST TIME

eighth ORDINAL

US GPE

his 20s DATE

Boston GPE

Boston GPE

Massachusetts GPE

Wuhan GPE

China GPE

Yesterday DATE

The Hal Turner Radio Show ORG

Presbyterian Hospital ORG

Pittsburgh GPE

Corona Virus ORG

FORCED Nurses, Doctors and Staff ORG

Non-Disclosure Agreements ORG

American NORP

Doctors, Nurses ORG

around 65 million CARDINAL

China GPE

the start of spring DATE

259 CARDINAL

China GPE

dozens CARDINAL

259 CARDINAL

The Medical Infrastructure of the United States ORG

ten days DATE

Medical Infrastructure COLLAPSE ORG

USA GPE

China GPE

America GPE

PANIC ORG

SHTF ORG

Preps PERSON

FedEx GPE

One CARDINAL

Two CARDINAL

N-95 PERSON

N-100 GPE

HERE-N-95 ORG

HERE N-100 LOC

3M 2091 DATE

Get Eye Protection ORG

Gloves Here Wear ORG

Centrum PERSON

Bible WORK\_OF\_ART

decades DATE

Quarantine PERSON

the 6 weeks DATE

Preps PRODUCT

SUNDAY 10:15 AM EST TIME

Chinese NORP

Sunday DATE

# CARDINAL

Shi Zhengli PERSON

Chinese NORP

the #Wuhan Institute of Virology ORG

People's Daily ORG

China GPE

February 2, 2020 DATE

China GPE

the Wuhan Virology ORG

100% PERCENT

China GPE

Virus ORG

Frankenstein PERSON

Virus ORG

Sorry China LOC

Today DATE

February 1 DATE

a couple more bucks MONEY

HAROLD PERSON

1906 DATE

Paterson Plank Road 1-A EVENT

North Bergen LOC

NJ 07047-1902 ORG

US Dollars ORG

two CARDINAL

a couple cents MONEY

Advertiser ORG

COVER ORG

COST ORG

one CARDINAL

Advertiser ORG

two to three cents MONEY

2005 DATE

79% PERCENT

Fauci PERSON

Birx PERSON

CDC ORG

America GPE

over 150,000 CARDINAL

nearly 80% PERCENT

US GPE

2020 RNC Night 2 Live-Stream Video Starting TIME

8 PM ET TIME

First ORDINAL

Melania Trump PERSON

State ORG

Mike Pompeo PERSON

Rand Paul PERSON

Mary Ann Mendoza PERSON

Eric Trump PERSON

Tiffany Trump PERSON

Nick Sandmann PERSON

100,000 CARDINAL

American NORP

Wednesday night TIME

Ramin Osoui PERSON

Laura Ingraham PERSON

2 billion CARDINAL

Fauci PERSON

Privacy Policy and Terms of Use You ORG

97% PERCENT

758 CARDINAL

3% PERCENT

20 CARDINAL

Votes ORG

Ramin Oskoui PERSON

tens of thousands CARDINAL

Oskoui PERSON

Fauci PERSON

Hahn PERSON

Redfield PERSON

Physicians NORP

Over 100,000 CARDINAL

Via The Ingraham Angle PERSON

CNN ORG

Monday DATE

the UN Intergovernmental Panel on Climate Change (IPCC ORG

1.5 degrees Celsius QUANTITY

2.7 degrees QUANTITY

Fahrenheit WORK OF ART

as early as 2030 DATE

hundreds of millions MONEY

today DATE

two-thirds CARDINAL

about 1 degree CARDINAL

the next few years DATE

1.5 degrees QUANTITY

Andrew King PERSON

the University of Melbourne ORG

45% PERCENT

2010 DATE

2030 DATE

zero CARDINAL

2050 DATE

around 1.5 degrees QUANTITY

1.5 degrees QUANTITY

the Paris Agreement LAW

NASA ORG

Operation IceBridge ORG

Greenland GPE

Arctic LOC

One CARDINAL

1 degree QUANTITY

Arctic LOC

Panmao Zhai PERSON

IPCC Working Group I. ORG

just below 1.5 degrees QUANTITY

summer DATE

Europe LOC

this summer DATE

3 degrees QUANTITY

between 70 and 90% PERCENT

Australia GPE

Great Barrier Reef FAC

1.5 degrees QUANTITY

Hans-Otto Pörtner PERSON

Co-Chair of IPCC Working Group II ORG

1.5 degrees QUANTITY

2 degrees QUANTITY

10 cm QUANTITY

2100 DATE

Arctic Ocean LOC

summer DATE

70% to 90% PERCENT

IPCC ORG

Monday DATE

three years DATE

2015 DATE

Paris Climate Agreement EVENT

Paris GPE

197 CARDINAL

2 degrees QUANTITY

1.5 degrees QUANTITY

C. NORP

The United States GPE

Donald Trump PERSON

a year and half later DATE

1.5 degree QUANTITY

Jim Skea PERSON

IPCC Working Group III ORG

Sarah Perkins-Kirkpatrick PERSON

the Climate Change Research Center ORG

the University of New South Wales ORG

The next few years DATE

One CARDINAL

two CARDINAL

Trump ORG

Today the world's WORK\_OF\_ART

US GPE

Al Gore PERSON

Trump ORG

American NORP

Barack Obama PERSON

US GPE

Nov. 28, 2020 DATE

CIA ORG

the People's Republic of China GPE

PRC GPE

The Criminal Complaint ORG

this morning TIME

John C. Demers PERSON

U.S. GPE

FBI ORG

Alan E. Kohler Jr. PERSON

Charge of the FBI's ORG

Honolulu Field Office ORG

Canadian NORP

Conservative Beaver ORG

Canadian NORP

Spanish LANGUAGE

Chinese NORP

American NORP

democratic NORP

John C. Demers PERSON

Chinese NORP

today DATE

Hawaii GPE

U.S. GPE

China GPE

the?People's Republic of China GPE

the United States GPE

Alan E. Kohler Jr. PERSON

FBI ORG

Counterintelligence Division ORG

FBI ORG

Chinese NORP

the United States GPE

years DATE

Eli Miranda PERSON

Charge of the FBI's ORG

Honolulu Division ORG

FBI ORG

Honolulu Division ORG

FBI ORG

Obama PERSON

Obama PERSON

Shane Smedley @ShaneSmedley ORG

November 30, 2020 DATE

Obama GPE

5 days DATE

ccp ORG

Self-Aware ORG

December 1, 2020 DATE

Google ORG

first ORDINAL

Beaumont GPE

Texas GPE

Jefferson County GPE

this morning TIME

one CARDINAL

the Beaumont Police Department ORG

48-year DATE

Henri Paul Johnson PERSON

sixteen hours TIME

52-year-old DATE

between 1400 to 1800 degrees DATE

Fahrenheit PERSON

Jenna Davis PERSON

one CARDINAL

Henri Johnson's ORG

about 15 seconds TIME

first ORDINAL

Davis PERSON

Johnson PERSON

Canada GPE

Canada GPE

Mike Adams of Natural News Natural News PERSON

Canadian NORP

Programmable Hydraulic Guillotines WORK OF ART

Canada GPE

COVID-19 ORG

45045-190091 DATE

45045-190091 DATE

National Capital Region Notice ORG

Programmable Hydraulic Guillotines WORK\_OF\_ART

Canadian NORP

Canada GPE

COVID-19 ORG

Guillotine ORG

Cambridge GPE

Antifa PERSON

BLM ORG

America GPE

the French Revolution EVENT

Canadian NORP

Canada GPE

COVID-19 ORG

Canada GPE

Canada GPE

Canadian NORP

Justin Trudeau PERSON

Canadian NORP

Nazi Germany GPE

years DATE

Ontario PERSON

Randy Hillier PERSON

Ontario GPE

Randy Hillier PERSON

Lanark PERSON

Kingston ORG

covid-19 PERSON

Hillier PERSON

Premier ORG

Premier ORG

Ontario GPE

Hillier PERSON

YouTube ORG

Canadian NORP

Guillotines Mandatory Requirements WORK\_OF\_ART

Guillotine PERSON

Guillotine ORG

Guillotine PERSON

Guillotine NORP

Guillotine PERSON

Guillotine PERSON

Guillotine PERSON

Guillotine ORG

a "1 year DATE

n't GPE

One CARDINAL

Canada GPE

Canada GPE

COVID-19 ORG

Cutting Machine WORK\_OF\_ART

France GPE

China GPE

Canada GPE

French LANGUAGE

China GPE

Canada GPE

Justin Trudeau PERSON

China GPE

Joe Biden PERSON

Chinese NORP

Western Canada ORG

Western Canada ORG

Communist NORP

America GPE

Dave Hodges PERSON

The Commonsense Show EVENT

Americans NORP

Canadian NORP

Chinese NORP

Steve Quayle PERSON

three CARDINAL

Communist NORP

China GPE

Canada GPE

the United States GPE

NATO ORG

American NORP

the Bush Sr. FAC

CSS ORG

UN ORG

America GPE

Harris-Biden ORG

Trudeau PERSON

Canada GPE

Canadian NORP

Canada GPE

Americans NORP

Earth LOC

the World Meteorological Organisation ORG

Cedro ORG

Quixadá GPE

Brazil GPE

2016 DATE

the hottest year DATE

2017 DATE

the World Meteorological Organisation ORG

WMO ORG

2016 DATE

Tuesday DATE

El Niño ORG

2016 DATE

US GPE

February DATE

El Niño ORG

2017 DATE

David Carlson PERSON

WMO ORG

Earth LOC

Jeffrey Kargel PERSON

the University of Arizona ORG

US GPE

Prof David Reay PERSON

the University of Edinburgh ORG

Donald Trump PERSON

Trump ORG

Republicans NORP

Congress ORG

Robert Watson PERSON

UK GPE

University of East Anglia ORG

UN ORG

Watson PERSON

Trump ORG

WMO ORG

Petteri Taalas PERSON

1880 DATE

about 115,000 years ago DATE

4m years DATE

2017 DATE

US GPE

February DATE

Australia GPE

Arctic ice LOC

October DATE

six consecutive months DATE

four-decade DATE

Prof Julienne Stroeve PERSON

University College London ORG

UK GPE

**Eel PERSON** 

Lowell's Cove, FAC

Maine GPE

US GPE

Emily Shuckburgh PERSON

the British Antarctic Survey ORG

Arctic LOC

Greenland GPE

Arctic sea LOC

Europe LOC

Asia LOC

North America LOC

between November 2014 and February 2016 DATE

El Niño ORG

15mm QUANTITY

five years DATE

recent decades DATE

2016 DATE

Arctic LOC

tens CARDINAL

Australia GPE

February DATE

Taalas GPE

the Mother Jones Daily ORG

Newt Gingrich PERSON

one CARDINAL

Donald Trump's PERSON

Cabinet ORG

Republican NORP

Trump ORG

Hillary Clinton PERSON

Gingrich PERSON

two years ago DATE

Trump ORG

four years ago DATE

Electoral College ORG

Election Day 2012 DATE

Mitt Romney PERSON

the Electoral College ORG

Barack Obama PERSON

Trump ORG

April DATE

Ben Carson PERSON

the Electoral College ORG

Gingrich PERSON

Trumpsters ORG

Trump ORG

2014 DATE

Gingrich PERSON

National Popular Vote ORG

11 CARDINAL

one CARDINAL

an additional 12 CARDINAL

Americans NORP

two years ago DATE

Gingrich PERSON

Gingrich PERSON

Americans NORP

Gingrich PERSON

314 million CARDINAL

Earth LOC

Gingrich PERSON

America GPE

Clinton PERSON

Trump ORG

60 million CARDINAL

59.8 million CARDINAL

one-fifth CARDINAL

Americans NORP

Gingrich PERSON

Trump ORG

one CARDINAL

the White House FAC

Trump ORG

2012 DATE

The United States GPE

More than 300 CARDINAL

between June and August DATE

under 18 DATE

796 CARDINAL

312 CARDINAL

under 18 DATE

the Press Association ORG

Prevent NORP

Channel ORG

last week DATE

14-year-old DATE

Britain GPE

Australia GPE

the National Police Chiefs' Council ORG

July DATE

349 CARDINAL

more than 10 CARDINAL

the previous month DATE

327 CARDINAL

120 CARDINAL

August DATE

the summer DATE

between June and August DATE

2012/13 CARDINAL

the first year DATE

England GPE

Wales GPE

the first three months DATE

2014/15 CARDINAL

Channel ORG

2007 DATE

one CARDINAL

five CARDINAL

Rashad Ali PERSON

the Institute for Strategic Dialogue ORG

later this year DATE

David Anderson QC PERSON

Britain GPE

Muslim NORP

last month DATE

14-year-old DATE

Muslim NORP

Isis ORG

London GPE

the North Pole LOC

Arctic LOC

winter DATE

nearly four decades DATE

the third straight year DATE

Arctic LOC

Arctic sea LOC

Mark Serreze PERSON

the National Snow and Ice Data Center ORG

Boulder GPE

Colo. GPE

Serreze PERSON

summer DATE

2030 DATE

a decade DATE

two CARDINAL

The Independent ORG

six CARDINAL

10 CARDINAL

Britain GPE

Chris Grayling PERSON

thousands CARDINAL

weeks DATE

Govia Thameslink Rail GTR ORG

last week DATE

East Coast LOC

Virgin Trains PERSON

Stagecoach PERSON

BMG Research ORG

last week DATE

64 per cent MONEY

just 19 per cent MONEY

56 per cent MONEY

only 15 per cent MONEY

Inside Politics ORG

**Brexit PERSON** 

Please ORG

Please ORG

Inside Politics ORG

# CARDINAL

The Independent ORG

Inside Politics ORG

**Brexit PERSON** 

Inside Politics ORG

# CARDINAL

The Independent ORG

80 per cent MONEY

Some 64 CARDINAL

only 20 CARDINAL

More than 15,000 CARDINAL

May DATE

GCSEs ORG

Cambridgeshire GPE

Hertfordshire GPE

Last week DATE

Mr Grayling ORG

recent years DATE

Andy McDonald PERSON

Network Rail ORG

Network Rail ORG

UK GPE

50 1 CARDINAL

UK GPE

UK GPE

13 April 2021 DATE

the early hours of Tuesday morning TIME

Bamburgh Lighthouse ORG

Northumberland GPE

UK GPE

12 April DATE

2021 DATE

Rebecca Richardson PERSON

Genevieve Florence PERSON

Clissold Leisure Centre ORG

London GPE

UK GPE

11 April 2021 DATE

Fox ORG

Hill PERSON

Denmark Hill PERSON

London GPE

EPA ORG

UK GPE

10 April 2021 DATE

The Death Gun Salute ORG

the Honourable Artillery Company ORG

Britain GPE

Prince Philip GPE

the The Tower of London AFP FAC

Getty UK ORG

9 April 2021 DATE

Buckingham Palace FAC

London GPE

Britain GPE

Prince Philip GPE

Elizabeth II's PERSON

Prince Philip PERSON

more than a month DATE

April 9 DATE

Buckingham Palace FAC

99 DATE

Getty Images ORG

UK GPE

8 April 2021 DATE

James King PERSON

4:05 TIME

REUTERS ORG

UK GPE

7 April 2021 DATE

the Independent Workers' Union of Great Britain ORG

Deliveroo GPE

London GPE

UK GPE

6 April 2021 DATE

Seaham Lighthouse ORG

Durham PA GPE

UK GPE

5 April 2021 DATE

Lusamba Katalay ORG

Belly Mujinga GPE

Victoria GPE

London GPE

first ORDINAL

Belly Mujinga GPE

London GPE

Victoria GPE

UK GPE

4 April 2021 DATE

Easter Sunday at EVENT

Hengistbury Head FAC

UK GPE

3 April 2021 DATE

Bill PERSON

London GPE

EPA ORG

UK GPE

2 April 2021 DATE

Bamburgh Croquet ORG

COVID-19 ORG

Northumberland GPE

Britain GPE

UK GPE

1 April 2021 DATE

St Nicholas' Park PERSON

Warwick GPE

UK GPE

this week DATE

Easter ORG

PA UK ORG

31 March 2021 DATE

the National Covid Memorial Wall ORG

London GPE

EPA ORG

UK GPE

30 March DATE

Jamie Klingler PERSON

Reclaim These Streets ORG

Clapham Common FAC

London GPE

the Chief Inspector of Constabulary ORG

Thomas Winsor PERSON

Metropolitan Police ORG

PA UK ORG

29 March 2021 DATE

the Mersey Bowman Lawn Tennis Club ORG

Liverpool GPE

England GPE

March 29, 2021 DATE

England GPE

third ORDINAL

up to six CARDINAL

England GPE

Monday DATE

second ORDINAL

AFP ORG

Getty Images ORG

UK GPE

28 March 2021 DATE

Tenby Lifeboat ORG

Tenby GPE

Wales Reuters ORG

UK GPE

27 March 2021 DATE

Manchester Reuters ORG

UK GPE

26 March 2021 DATE

Andy Warhol PERSON

Elizabeth II PERSON

Reigning Queens ORG

1985 DATE

Elizabeth II PERSON

1985 DATE

Christie's ORG

London GPE

UK GPE

25 March 2021 DATE

Charlton Lido PERSON

London GPE

PA UK ORG

24 March 2021 DATE

Scottish Green Party ORG

Patrick Harvie PERSON

Lorna Slater PERSON

Edinburgh GPE

PA GPE

UK GPE

23 March 2021 DATE

Westbourne House ORG

Sheffield GPE

the National Reflection day EVENT

first ORDINAL

first ORDINAL

UK GPE

Tom Maddick PERSON

UK GPE

22 March 2021 DATE

Couple CARDINAL

Sean Henry PERSON

Newbiggin GPE

the-Sea LOC

Northumberland GPE

UK GPE

21 March 2021 DATE

Jersey GPE

meteor ORG

UK GPE

20 March 2021 DATE

London GPE

UK GPE

19 March 2021 DATE

St. James's Park GPE

London GPE

Britain GPE

UK GPE

18 March 2021 DATE

Highland PERSON

the National Trust's ORG

Cambridgeshire GPE

UK GPE

17 March 2021 DATE

early in the morning TIME

PA GPE

UK GPE

16 March 2021 DATE

60-metre QUANTITY

12-year-old DATE

Ansha PERSON Ethiopia GPE WaterAid ORG Whitby Beach GPE Yorkshire GPE hours TIME each day DATE UK GPE 15 March 2021 DATE 2021 DATE Teresa Solar ORG Exchange Flags ORG one CARDINAL 11th ORDINAL Liverpool Biennial ORG March 20, 2021 DATE nine CARDINAL PA Photo ORG Monday March 15, 2021 DATE PA Photo ORG second ORDINAL UK GPE 14 March 2021 DATE Sarah Everard PERSON Clapham Common FAC Sarah Everard's PERSON Parliament Square FAC London AFP PERSON Getty UK ORG 13 March 2021 DATE Sarah Everard PERSON Clapham Common FAC London GPE Getty Images ORG UK GPE 12 March DATE 2021 DATE Angela Rayner PERSON the Labour Party ORG Jonathan Ashworth PERSON Shadow Health WORK\_OF\_ART London GPE 1 percent PERCENT Thursday DATE May PA DATE UK GPE 11 March 2021 DATE Newhaven GPE England AFP/ PERSON UK GPE

10 March 2021 DATE

Great Chart Golf FAC

Leisure in Ashford ORG

UK GPE

9 March 2021 DATE

Kielder Water FAC

Northumberland GPE

the Border Mires ORG

50 year old DATE

the Border Mires in WORK\_OF\_ART

Northumberland GPE

UK GPE

8 March 2021 DATE

the Jewellery Quarter Academy ORG

Birmingham GPE

the West Midlands LOC

England GPE

first ORDINAL

two months DATE

first ORDINAL

UK GPE

7 March 2021 DATE

NHS ORG

NHS Workers ORG

Unite ORG

Downing Street FAC

London GPE

1% PERCENT

UK GPE

6 March 2021 DATE

Scottish NORP

Ibrox Stadium FAC

UK GPE

March 2021 DATE

Sutton Manor GPE

St Helens PERSON

a clear DATE

Spring day DATE

PA GPE

UK GPE

4 March 2021 DATE

Health ORG

Matt Hancock PERSON

the Centre for Virus Research ORG

Glasgow PA GPE

UK GPE

March 2021 DATE

Exchequer ORG

Rishi Sunak ORG

11 Downing Street ORG

London GPE

the House of Commons ORG

Budget PA UK ORG

2 March 2021 DATE

the Carbis Bay Estate PRODUCT

G7 CARDINAL

Getty UK ORG

1 March 2021 DATE

A J Thompson & Sons ORG

the Romney Marsh ORG

Kent PERSON

the first day DATE

UK GPE

28 CARDINAL

February 2021 DATE

Scarborough GPE

North Yorkshire GPE

UK GPE

27 February 2021 DATE

Tom Moore PERSON

the Armed Forces at Bedford Crematorium ORG

Bedford GPE

EPA ORG

UK GPE

26 February 2021 DATE

New Brighton GPE

Wirral PA UK ORG

25 February 2021 DATE

Shadow PRODUCT

Anneliese NORP

the Institute of Global Prosperity ORG

Labour Party ORG

London GPE

UK GPE

24 February 2021 DATE

Britain GPE

Health ORG

Matt Hancock PERSON

Downing Street FAC

London GPE

Britain GPE

UK GPE

23 CARDINAL

February 2021 DATE

Two CARDINAL

Boscombe GPE

Dorset PA RMT GPE

Mr Grayling ORG

Tory NORP

Network Rail ORG

GTR ORG

Network Rail ORG

Mark Carne PERSON

Queen PERSON

The Department of Transport ORG

BMG ORG

1,490 CARDINAL

Great Britain GPE

5th ORDINAL

8th June DATE

BMG ORG

the British Polling Council ORG

Democrats NORP

today DATE

Social Security ORG

Democrats NORP

years DATE

a couple of years DATE

Social Security ORG

a decade DATE

the present 65 years old DATE

75 years old DATE

Biden PERSON

Social Security ORG

Joe Biden PERSON

The United States Of America GPE

Democrats NORP

Americans NORP

America GPE

Mexico GPE

Tijuana GPE

One CARDINAL

Anna PERSON

Elsa PERSON

Disney ORG

Frozen PERSON

camera Sunday DATE

Reuters ORG

Kim Kyung-Hoon PERSON

Central American NORP

San Diego GPE

U.S. GPE

Kim PERSON

NBC News ORG

Monday DATE

two CARDINAL

One CARDINAL

NBC News ORG

Maria Mesa PERSON

39-year-old DATE

five CARDINAL

Honduras GPE

the Benito Juarez Shelter FAC

Tijuana GPE

Mesa ORG

5-year-old DATE

Saira PERSON

Mesa ORG

3-year-old DATE

James PERSON

One CARDINAL

Kim PERSON

Mesa ORG

five CARDINAL

Louisiana GPE

Kim PERSON

NBC News ORG

Donald Trump PERSON

Trump ORG

Mexico GPE

Migrants PRODUCT

U.S.A. GPE

Border ORG

Congress ORG

WALL ORG

Donald J. Trump PERSON

@realDonaldTrump ORG

November 26, 2018 DATE

Kim PERSON

44-year-old DATE

South Korean NORP

Tokyo GPE

the 16 years DATE

Reuters ORG

2011 DATE

Japan GPE

Kim PERSON

Nov. 14 DATE

Mexico City GPE

American NORP

Sunday DATE

Kim PERSON

El Chaparral ORG

U.S. GPE

Reuters ORG

Kim Kyung-Hoon PERSON

Minamisoma GPE

Japan GPE

April 11, 2011 DATE

Minamisoma PERSON

11 miles QUANTITY

Fukushima PERSON

Kim Kyung-Hoon PERSON

Mexican NORP

about 20 meters QUANTITY

Kim PERSON

a few seconds TIME

Kim PERSON

two CARDINAL

One CARDINAL

Tijuana GPE

Mexicali GPE

Mexico GPE

Nov. 20 DATE

Kim Kyung-Hoon PERSON

Kim PERSON

Kim PERSON

CORRECTION PERSON

Nov. 27, 2018 DATE

7:50 a.m. TIME

One CARDINAL

Gabe Gutierrez PERSON

Tijuana GPE

Corky Siemaszko PERSON

New York City GPE

Rick Egan PERSON

The Salt Lake Tribune ORG

Mitt Romney PERSON

United States GPE

Senate ORG

the State Capitol FAC

Thursday, March 15, 2018 DATE

Mitt Romney's PERSON

2012 DATE

GOP ORG

Utah GPE

U.S. Senate ORG

\$1 million MONEY

Senate ORG

Tuesday DATE

a few hundred thousand dollars MONEY

vear-end DATE

\$1.3 million MONEY

\$1 million MONEY

Romney PERSON

\$1.68 million MONEY

April 1 this year DATE

Senate ORG

526,516 MONEY

\$1.15 million MONEY

\$1 million MONEY

Mitt Romney PERSON

Jenny Wilson PERSON

Democratic NORP

Utahns ORG

Romney PERSON

Mitt PERSON

Utah GPE

Wilson ORG

another week DATE

the Utah Democratic Convention ORG

a week later DATE

GOP ORG

Romney PERSON

Larry Meyers PERSON

St. George GPE

one CARDINAL

11 CARDINAL

Republicans NORP

Romney PERSON

April 21 DATE

Utah GPE

about \$15,000 MONEY

about \$4,000 MONEY

Romney PERSON

4,000 CARDINAL

one CARDINAL

Romney PERSON

Mike Kennedy PERSON

Romney PERSON

\$1 million MONEY

millions of dollars MONEY

Utah GPE

Barack] Obama PERSON

Kennedy PERSON

251,143 MONEY

289,000 MONEY

257,488 MONEY

31,535 MONEY

Senate ORG

Tuesday DATE

the Federal Election Commission's ORG

first ORDINAL

Senate ORG

FEC ORG

Tuesday DATE

afternoon TIME

Spencer Cox PERSON

24,478 CARDINAL

28,000 CARDINAL

Romney PERSON

Monday DATE

Romney PERSON

Republican NORP

U.S. Senate ORG

Romney PERSON

Romney PERSON

Romney PERSON

60 percent PERCENT

Republicans NORP

Nov. 6 DATE

60 percent PERCENT

two CARDINAL

Romney PERSON

Romney PERSON

George W. Bush PERSON

8,100 MONEY

8,100 MONEY

Richard Marriott PERSON

Host Hotels ORG

Harris Simmons PERSON

Zions Bancorp ORG

John Cumming PERSON

Powdr Corp. ORG

GETTY - STOCK IMAGE Official ORG

EU ORG

nine CARDINAL

Britain GPE

SUBSCRIBE Invalid PRODUCT

last night TIME

six years DATE

2010 DATE

some 2.2 million CARDINAL

Britain GPE

28.2 million CARDINAL

just 1.1 million CARDINAL

Europe LOC

last year DATE

3.4 million CARDINAL

30.3 million CARDINAL

2.2 million CARDINAL

EU ORG

1.2 million CARDINAL

Anti-EU ORG

vears DATE

Britain GPE

last year's DATE

Steven Woolfe PERSON

EU ORG

Theresa May PERSON

British NORP

six years DATE

EU ORG

Brexit PERSON

28 CARDINAL

the European Union ORG

Getty 1 PERSON

29 CARDINAL

the European Union A ORG

the Office for National Statistics ORG

11.2 per cent MONEY

UK GPE

2016 DATE

EU ORG

Steven Woolfe PERSON

761,000 CARDINAL

508,000 CARDINAL

EU ORG

669,000 CARDINAL

510,000 CARDINAL

EU ORG

Eight per cent MONEY

eight CARDINAL

European NORP

EU8 FAC

EU ORG

2004 DATE

Steven Woolfe PERSON

Czech Republic GPE

Estonia GPE

Hungary GPE

Latvia GPE

Lithuania GPE

Poland GPE

Slovakia GPE

Slovenia GPE

EU ORG

11 per cent MONEY

Seven per cent MONEY

eight CARDINAL

EU ORG

One CARDINAL

eight CARDINAL

382,000 CARDINAL

EU ORG

Ukip GPE

Jane Collins PERSON

EU ORG

Brussels GPE

Some 701,000 CARDINAL

UK GPE

more than a quarter DATE

EU ORG

Yesterday DATE

tens of thousands CARDINAL

European NORP

Britain GPE

Theresa May PERSON

UK GPE

Kemi Olunloye PERSON

the United States GPE

Muhammadu Buhari PERSON

Abba Kyari PERSON

Abba Kyari PERSON

Coronavirus GPE

March DATE

Lagos GPE

Thursday DATE

Olunloye PERSON

Buhari PERSON

Cabinet ORG

# CARDINAL

⊕ ORG

# CARDINAL

Abba Kyari PERSON

daily DATE

Osagie Ehanire PERSON

Kyari PERSON

Nigerian NORP

Ehanire ORG

SGF ORG

UK GPE

today DATE

first ORDINAL

a month ago DATE

Today DATE

Nicola ORG

CBE ORG

Chris Snowden PERSON

Julia Goodfellow PERSON

first ORDINAL

around 100 years DATE

UUK ORG

the last 5 years DATE

David Willetts PERSON

**Greg Clark PERSON** 

OECD ORG

Graduate Labour Market ORG

2007 DATE

87.5% PERCENT

```
UK GPE
HE GPE
4 CARDINAL
10 CARDINAL
Unfinished PERSON
the coming months DATE
3 CARDINAL
firstly ORDINAL
2020 DATE
2009 DATE
secondly ORDINAL
first ORDINAL
2012 DATE
HEPI ORG
BBC/ComRes ORG
only around half CARDINAL
the Small Business, Enterprise and Employment Act ORG
March DATE
Department for Business, Innovation and Skills BIS and ORG
Department for Education DfE ORG
Department for Work and Pensions DWP ORG
National Student Survey ORG
US GPE
the Competition and Markets Authority ORG
Higher Education Policy Institute Hepi ORG
as many as three-quarters CARDINAL
the last 3 years DATE
1992 DATE
Between 2006 and 2015 DATE
around 55% PERCENT
around 45% PERCENT
22,000 MONEY
Last year DATE
CBI /Pearson Education ORG
Skills ORG
47% PERCENT
Excellence Framework PERSON
Parliament ORG
David Willetts PERSON
the Times Higher Education ORG
English LANGUAGE
the Research Excellence Framework ORG
the Teaching Excellence Framework ORG
TEF ORG
TEF ORG
TEF ORG
REF ORG
the Higher Education Funding Council for England ORG
the Quality Assurance Agency ORG
2010 DATE
```

```
the Department of Business, Innovation ORG
Skills PRODUCT
the autumn DATE
One CARDINAL
TEF ORG
UK GPE
Last week DATE
2015 DATE
the Higher Education Achievement Report ORG
2:1 degrees QUANTITY
2:1s CARDINAL
UK GPE
300% PERCENT
the 1990s DATE
Over 70% PERCENT
First ORDINAL
2:1 CARDINAL
7 CARDINAL
the past 5 years DATE
just 47% PERCENT
the mid-1990s DATE
2013 DATE
14 CARDINAL
over 50% PERCENT
second ORDINAL
The Higher Education Academy ORG
nearly half CARDINAL
The Teaching Excellence Framework PERSON
the Teaching Excellence Framework ORG
First ORDINAL
2:1 CARDINAL
2:2 CARDINAL
Third ORDINAL
GPA ORG
13 CARDINAL
2:1 CARDINAL
the Teaching Excellence Framework ORG
Student Academic Experience Survey ORG
30.5 hours TIME
14 hours TIME
weekly DATE
Mass Communications and Documentation ORG
as little as 22 hours QUANTITY
GPA ORG
Robert Burgess PERSON
the Grade Point Average Advisory Group ORG
UK GPE
the end of the decade DATE
GPA ORG
```

Higher Education Academy ORG

2012 DATE

this academic year DATE

2020 CARDINAL

2009 DATE

Les Ebdon PERSON

the Office of Fair Access ORG

Degree Apprenticeships PERSON

2-year DATE

the National Strategy for Access and Student Success ORG

their college years DATE

CNN ORG

Sweden GPE

Sweden GPE

first ORDINAL

two CARDINAL

Sweden GPE

Nicolaigarden GPE

Egalia GPE

Latin LANGUAGE

first ORDINAL

Egalia GPE

the age of 10 DATE

The Global Early Adolescent Study ORG

15 CARDINAL

between the ages of 10 DATE

Sweden GPE

fourth ORDINAL

the Education Act LAW

1998 DATE

Lotta Rajalin PERSON

first ORDINAL

one to CARDINAL

five-year-olds DATE

Stockholm GPE

Rajalin ORG

heroines CARDINAL

Bangladesh GPE

monsoon season DATE

Bangladesh GPE

almost one third CARDINAL

Shidhulai Swanirvar Sangstha PERSON

Each morning TIME

up to CARDINAL

30 CARDINAL

hundreds CARDINAL

10 CARDINAL

UK GPE

Sevenoaks School ORG

UK GPE

VR ORG

three CARDINAL

10 CARDINAL

Bali PERSON

the Green School ORG

Bali GPE

The Green School ORG

three CARDINAL

3 CARDINAL

10 CARDINAL

Ghana GPE

Ghana GPE

over 40 hours TIME

10 CARDINAL

Canada GPE

Canada GPE

U.S. GPE

iPads ORG

Google Classroom ORG

10 CARDINAL

Netherlands GPE

Steve Jobs Schools PERSON

Apple ORG

iPad ORG

Netherlands GPE

Johannesburg GPE

2016 DATE

10 CARDINAL

Sweden GPE

Egalia GPE

Stockholm GPE

Sweden GPE

10 CARDINAL

Singapore GPE

Pepper ORG

last year DATE

Singapore GPE

Pepper ORG

10 CARDINAL

Denmark GPE

Denmark GPE

the Danish Forest and Nature Agency ORG

over 10 percent PERCENT

Danish NORP

10 CARDINAL

USA GPE

New York GPE

2018-19 school year DATE

India GPE

Botswana GPE

Japan GPE

Spain GPE

Rajalin PERSON

CNN ORG

Lotta Rajalin PERSON

Nicolaigarden ORG

Egalia GPE

Rajalin PERSON

two CARDINAL

Rajalin PERSON

Rajalin PERSON

Ben Kenward PERSON

Uppsala University ORG

Sweden GPE

Oxford Brookes University ORG

England GPE

Swedish NORP

'Brainwashing children' However WORK\_OF\_ART

the years DATE

Swedish NORP

David Eberhard PERSON

CNN ORG

Eberhard PERSON

Rajalin PERSON

Rajalin PERSON

Eberhard PERSON

Sweden GPE

Rajalin PERSON

Philip Hwang PERSON

the University of Gothenburg ORG

Rajalin PERSON

CNN ORG

GETTY - STOCK IMAGE Official ORG

EU ORG

nine CARDINAL

Britain GPE

SUBSCRIBE Invalid PRODUCT

last night TIME

six years DATE

2010 DATE

some 2.2 million CARDINAL

Britain GPE

28.2 million CARDINAL

just 1.1 million CARDINAL

Europe LOC

last year DATE

3.4 million CARDINAL

30.3 million CARDINAL

2.2 million CARDINAL

EU ORG

1.2 million CARDINAL

Anti-EU ORG

years DATE

Britain GPE

last year's DATE

Steven Woolfe PERSON

EU ORG

Theresa May PERSON

British NORP

six years DATE

EU ORG

**Brexit PERSON** 

28 CARDINAL

the European Union ORG

Getty 1 PERSON

29 CARDINAL

the European Union A ORG

the Office for National Statistics ORG

11.2 per cent MONEY

UK GPE

2016 DATE

EU ORG

Steven Woolfe PERSON

761,000 CARDINAL

508,000 CARDINAL

EU ORG

669,000 CARDINAL

510,000 CARDINAL

EU ORG

Eight per cent MONEY

eight CARDINAL

European NORP

EU8 FAC

EU ORG

2004 DATE

Steven Woolfe PERSON

Czech Republic GPE

Estonia GPE

Hungary GPE

Latvia GPE

Lithuania GPE

Poland GPE

Slovakia GPE

Slovenia GPE

EU ORG

11 per cent MONEY

Seven per cent MONEY

eight CARDINAL

EU ORG

One CARDINAL

eight CARDINAL

382,000 CARDINAL

EU ORG

Ukip GPE

Jane Collins PERSON

EU ORG

Brussels GPE

Some 701,000 CARDINAL

UK GPE

more than a quarter DATE

EU ORG

Yesterday DATE

tens of thousands CARDINAL

European NORP

Britain GPE

Theresa May PERSON

UK GPE

Three CARDINAL

four CARDINAL

today DATE

almost half CARDINAL

two CARDINAL

165 CARDINAL

the Royal College of Obstetricians ORG

Gynaecologists NORP

today DATE

almost half CARDINAL

30 minutes TIME

More than half CARDINAL

hours TIME

third ORDINAL

165 CARDINAL

NHS ORG

82 per cent MONEY

Just over 73 CARDINAL

weekday DATE

overnight TIME

10pm to 8am TIME

43 per cent MONEY

89 per cent MONEY

evening TIME

England NHS ORG

Birmingham GPE

Central Manchester University ORG

day DATE

all week DATE

Maureen Treadwell PERSON

the Birth Trauma Association ORG

night TIME

every year DATE

Jonathan Ashworth PERSON

Labour ORG

five-day DATE

Jonathan Ashworth PERSON

Labour ORG

day DATE

night TIME

Tories NORP

seven-day DATE

NHS ORG

five-day DATE

NHS ORG

Elizabeth Duff PERSON

the National Childbirth Trust ORG

2009 DATE

Imperial College London ORG

last year DATE

the weekend DATE

7 per cent MONEY

Cambridge University ORG

2010 DATE

overnight TIME

Scottish NORP

third ORDINAL

Angela Owens PERSON

Ella ORG

today DATE

Angela Owens PERSON

Ella ORG

today DATE

two weeks DATE

Saturday DATE

afternoon TIME

four days DATE

Christmas 2013 DATE

31-year-old DATE

Warrington Hospital ORG

GP ORG

Ella ORG

11pm TIME

Saturday DATE

two CARDINAL

Paul PERSON

Ella ORG

one CARDINAL

ten CARDINAL

Warrington Hospital ORG

12-month DATE

English LANGUAGE

NHS ORG

almost 40 per cent MONEY

2014/15 CARDINAL

24 per cent MONEY

1989 DATE

NHS England ORG

NHS ORG

last year DATE

overnight TIME

hours TIME

Tomorrow DATE

NHS ORG

first ORDINAL

Ofsted ORG

Tesco ORG

59,400 CARDINAL

last year DATE

nearly 119 million CARDINAL

Tesco ORG

4% PERCENT

2015 DATE

one CARDINAL

100 CARDINAL

Tesco ORG

the last financial year DATE

annual DATE

Tesco ORG

Dave Lewis PERSON

today DATE

Tesco ORG

UK GPE

nearly three years DATE

UK GPE

Tesco ORG

2013 DATE

last year DATE

Tesco ORG

2% PERCENT

0.5kg QUANTITY

Lewis PERSON

Unilever ORG

Tesco ORG

2014 DATE

the Global Summit of the Consumer Goods Forum ORG

Tesco ORG

Perfectly Imperfect WORK\_OF\_ART

February DATE

the end of 2017 DATE

Tesco ORG

this year DATE

future years DATE

millions CARDINAL

Tesco ORG

Dave Lewis PERSON

Neil Hall/REUTERS PERSON

This year DATE

Tesco ORG

Asda ORG

Morrisons PERSON

fifth ORDINAL

the next decade DATE

Coca-Cola ORG

Nestle ORG

the Courtauld Commitment ORG

2025 DATE

the Waste and Resources Action Programme Wrap ORG

1.9m tonnes QUANTITY

UK GPE

every year DATE

Tesco ORG

Richard Swannell PERSON

Wrap ORG

Tesco ORG

UK GPE

2025 DATE

200,000 tonne QUANTITY

UK GPE

WRAP ORG

27-plus-million CARDINAL

UK GPE

2025 DATE

163 CARDINAL

George Soros PERSON

Soros PERSON

Craigslist NORP

Paid Liberal Trolls of America ORG

6 hours TIME

30 MONEY

40 MONEY

70 cents MONEY

1 MONEY

Soros PERSON

Joe Barron PERSON

George Soros PERSON

500 MONEY

Art Tubolls PERSON

163 CARDINAL

Soros PERSON

Deluth GPE

Charlie Kirk PERSON

Turning Point PERSON

Turning Point PERSON

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an honest day DATE
NPR ORG
NPR ORG
NPR ORG
NPR ORG
Agree and Continue WORK_OF_ART
NPR ORG
CNN ORG
the United States GPE
the Centers for Disease Control and Prevention ORG
Jeff Lancashire PERSON
the National Center for Health Statistics ORG
Friday DATE
US GPE
US GPE
daily DATE
daily DATE
CDC ORG
Amazon ORG
Orwell ORG
Nineteen Eighty-Four DATE
the end of last week DATE
Trump ORG
Kellyanne Conway PERSON
Jezebel PERSON
Hannah Arendt PERSON
The Origins of Totalitarianism WORK OF ART
1951 DATE
Donald Trump PERSON
November DATE
Jews NORP
first ORDINAL
Nazi NORP
Trump ORG
anti-Jewish NORP
Europe LOC
the 19th century DATE
John Naughton's PERSON
David Runciman's PERSON
Talking Politics ORG
20 years DATE
Trump ORG
Washington GPE
Pennsylvania Avenue FAC
the Women's March EVENT
January DATE
Joshua Lott/AFP/Getty Images PERSON
Two CARDINAL
First ORDINAL
```

Antisemites NORP

Second ORDINAL

Runciman PERSON

one CARDINAL

Arendt ORG

Griselda PERSON

Arendt GPE

two CARDINAL

Arendt ORG

Germany GPE

1906 DATE

1933 DATE

Palestine GPE

Jewish NORP

Hitler PERSON

1964 DATE

Nazis NORP

Germans NORP

1933 DATE

Jewish NORP

Arendt ORG

Jewish NORP

around one CARDINAL

Paris GPE

German NORP

1937 DATE

New York GPE

1941 DATE

Gurs FAC

France GPE

Holocaust EVENT

Auschwitz ORG

Arendt ORG

New York GPE

1972 DATE

New York Times Co/Getty Images However ORG

the 1960s DATE

Eichmann NORP

Jerusalem GPE

Adolf Eichmann's PERSON

the Final Solution ORG

Jews NORP

Jews NORP

Eichmann PERSON

three minutes TIME

Martin Heidegger PERSON

the 20s DATE

Nazi NORP

Nazi NORP

The Origins of Totalitarianism and the US's WORK\_OF\_ART

today DATE

Islamophobia ORG

Arendt ORG

Origins PERSON

1958 DATE

The Human Condition ORG

Holocaust EVENT

Arendt ORG

two CARDINAL

First ORDINAL

Greek NORP

Greek NORP

Greeks PERSON

Arendt ORG

one CARDINAL

Trump ORG

Leninist ORG

Trump ORG

Mark Davis PERSON

the Bauman Institute ORG

Leeds GPE

1970 DATE

Downing Street FAC

January DATE

Ben Stansall/AFP/Getty Images Pollock PERSON

one CARDINAL

First ORDINAL

Muslims NORP

today DATE

Hannah Arendt PERSON

Downing Street FAC

Davis PERSON

one CARDINAL

one CARDINAL

Arendt ORG

The Twilight Zone WORK OF ART

NASA ORG

Antarctica LOC

the early 1960s DATE

sci-fi TV ORG

the Big Bang GPE

the Daily Star ORG

NASA ORG

Antarctic Impulsive Transient Antenna PRODUCT

ANITA ORG

Antarctica LOC

Earth LOC

Earth LOC

Manoa GPE

ANITA ORG

Earth LOC

the Daily Star ORG

ANITA ORG

Peter Gorham PERSON

the University of Hawaii ORG

Earth LOC

Gorham PERSON

Cornell University ORG

New Scientist ORG

the Big Bang ORG

13.8 billion years ago DATE

two CARDINAL

Ibrahim Safa PERSON

Catholic NORP

Kentucky GPE

Native American NORP

Vietnam GPE

Saturday DATE

Make America Great Again WORK OF ART

Covington Catholic High School ORG

Washington GPE

Friday DATE

Nathan Phillips PERSON

the Native American NORP

One CARDINAL

Phillips ORG

Donald Trump PERSON

MAGA ORG

Phillips ORG

Friday DATE

Alyssa Milano PERSON

Chris Evans PERSON

Diocese NORP

Nathan Phillips ORG

Native Americans NORP

Phillips PERSON

Church ORG

Covington ORG

Joe Meyer PERSON

Democrat NORP

64-year-old DATE

Phillips ORG

Nebraska GPE

Omaha GPE

The Washington Post ORG

Shoshana PERSON

nearly four years ago DATE

Phillips ORG

Phillips ORG

annual DATE

Arlington National Cemetery ORG

Native American NORP

Native American NORP

Standing Rock ORG

2016-2017 DATE

North Dakota GPE

Democratic NORP

US GPE

Deb Haaland PERSON

New Mexico's GPE

Laguna Pueblo PERSON

Twitter PERSON

Phillips ORG

Reuters ORG

Associated Press ORG

Britain GPE

0.7 per cent MONEY

last year DATE

£11.7 billion MONEY

£1 billion MONEY

the next two years DATE

the European Union ORG

yesterday DATE

The Department for International Development Dfid ORG

Britain GPE

EU ORG

UK GPE

second ORDINAL

vesterday DATE

Britain GPE

0.7 per cent MONEY

last year DATE

£11.7 billion MONEY

Jonathan Ashworth PERSON

Labour's Shadow Health ORG

today DATE

3-4 months DATE

Labour ORG

Britain GPE

3-4 months DATE

Labour ORG

an additional £25 million MONEY

Labour ORG

National Child Health Fund ORG

Britain GPE

Today DATE

Labour ORG

Jonathan Ashworth PERSON

Health Visitor Programme ORG

2015 DATE

Between 2016/17 CARDINAL

2017/18 DATE

£55.2 million MONEY

0 CARDINAL

this Tory Government ORG

8,244 CARDINAL

NHS ORG

December 2017 DATE

August 2013 DATE

10,309 CARDINAL

October 2015 DATE

more than 2,065 CARDINAL

20% PERCENT

just over two years DATE

12% PERCENT

a New Birth Visit ORG

17% PERCENT

the South West LOC

England GPE

17% PERCENT

6-8 week DATE

London GPE

1/3 CARDINAL

25% PERCENT

one year DATE

12 CARDINAL

London GPE

44% PERCENT

90% PERCENT

North East LOC

two and a half year DATE

just 64% PERCENT

London GPE

73% PERCENT

the East of England LOC

Jonathan Ashworth PERSON

today DATE

Infant Feeding PERSON

a Labour Government ORG

the Coalition Government ORG

UNICEF ORG

Scotland GPE

Northern Ireland GPE

70% PERCENT

the year 2030 DATE

Jonathan Ashworth PERSON

Labour's Shadow Health ORG

London GPE

Michael Marmot PERSON

UK GPE

70th ORDINAL

NHS ORG

David Cameron PERSON

Theresa May PERSON

more than 20 percent PERCENT

just over two years DATE

England GPE

UK GPE

Health Visitors ORG

3-4 months DATE

£25 million MONEY

National Child Health Fund ORG

20 per cent MONEY

100 per cent MONEY

Rainsy PERSON

Cambodia GPE

Sam Rainsy C PERSON

L GPE

seven CARDINAL

CNRP ORG

Cambodia GPE

Sam Rainsy C PERSON

L GPE

seven CARDINAL

CNRP ORG

Facebook Government ORG

the Phnom Penh Municipal Court's ORG

Sam Rainsy PERSON

eight CARDINAL

Ministry of Justice ORG

Chin Malin PERSON

yesterday DATE

Duch Sok Sarin PERSON

Monday DATE

Rainsy PERSON

25 years DATE

Mu Sochua PERSON

Eng Chhai Eang PERSON

22 years DATE

six CARDINAL

20 years DATE

Rainsy PERSON

Tioulong Saumura PERSON

Ou Chanrith PERSON

Ho Vann PERSON

Men Thavarin GPE

Nuth Romduol ORG

Kingdom GPE

November 9, 2019 DATE

Rainsy PERSON

Cambodian NORP

Hun Sen PERSON

Cambodian NORP

Rainsy PERSON

Hun Sen PERSON

Cambodia GPE

CNRP ORG

Malin PERSON

Malin PERSON

CNRP ORG

Cambodia GPE

130 CARDINAL

the past year DATE

nine CARDINAL

CNRP ORG

The US Embassy ORG

Phnom Penh GPE

yesterday DATE

nine CARDINAL

Cambodian NORP

American NORP

more than 130 CARDINAL

Cambodia GPE

2022 DATE

Cambodian People's Party ORG

Sok Eysan PERSON

yesterday DATE

CNRP ORG

2022 DATE

2023 CARDINAL

Commune and National Elections ORG

America GPE

Christian NORP

Americans NORP

Christianity NORP

Nazi Germany GPE

New Zealand GPE

Donald Trump PERSON

Executive Order PERSON

66 CARDINAL

today DATE

the United States of America GPE

IRS ORG

all these years DATE

Americans NORP

America GPE

Trump ORG

Executive Order PERSON

Kafka PERSON

Kafka PERSON

Kafka PERSON thirty years DATE Today DATE hundreds of millions MONEY two CARDINAL this late date DATE 1988 DATE the past thirty years DATE the previous two centuries DATE another election year DATE Netflix GPE one CARDINAL the Republican Party's ORG more than two CARDINAL less than two CARDINAL the past three decades DATE zero CARDINAL the next three decades DATE I.P.C.C. GPE last month DATE Scientific American NORP ten thousand CARDINAL the coming century DATE two degrees QUANTITY at least two CARDINAL first ORDINAL Nature GPE thousands CARDINAL New York City GPE Texans NORP Kafkaesque NORP the European Union's ORG Indonesia GPE American NORP China GPE COVID-19 ORG Officially GPE China GPE nearly 82,000 CARDINAL around 3,300 CARDINAL each day DATE China GPE Wuhan GPE 11 million CARDINAL the United States GPE nearly 500,000 CARDINAL more than 18,000 CARDINAL China GPE China GPE nearly 21 million CARDINAL

January DATE

February DATE

China Mobile Ltd. ORG

nearly 7.25 million CARDINAL

China GPE

Hong Kong Ltd. ORG

7.8 million CARDINAL

January and February DATE

China Telecom Corp. ORG

5.6 million CARDINAL

February DATE

the Associated Press ORG

China GPE

Tang Jinyuan PERSON

China GPE

the Epoch Times ORG

China GPE

Samantha Hoffman PERSON

the Australian Strategic Policy Institute ORG

Hoffman PERSON

the Associated Press ORG

Lunar New Year holidays EVENT

Bloomberg PERSON

Hoffman PERSON

China GPE

U.S. GPE

China GPE

the New York Times ORG

Fortune Magazine ORG

Hoffman PERSON

China GPE

China GPE

Hoffman PERSON

the United States GPE

Chinese NORP

Business Insider ORG

Beijing GPE

Hoffman PERSON

China GPE

Chinese Communist] Party ORG

Hoffman PERSON

Hoffman PERSON

China GPE

ten CARDINAL

China GPE

Hoffman PERSON

U.S. GPE

U.S. GPE

State ORG

Michael Pompeo PERSON

China GPE

Mike Pence PERSON

CNN ORG

America GPE

Hoffman PERSON

China GPE

Hoffman PERSON

Chinese Communist] Party ORG

YUKON DELTA NATIONAL WILDLIFE REFUGE ORG

Alaska GPE

Arctic LOC

Woods Hole Research Center ORG

a few feet QUANTITY

tens or even hundreds CARDINAL

the atmosphere centuries ago DATE

as much as 1.7 degrees QUANTITY

Fahrenheit WORK OF ART

the next several centuries DATE

Alaska GPE

350 miles QUANTITY

the Arctic Circle LOC

two CARDINAL

Yukon ORG

Kuskokwim PERSON

three feet QUANTITY

less than half CARDINAL

Max Holmes PERSON

Alaska GPE

Sarah Ludwig PERSON

Woods Hole ORG

Laura Jardine PERSON

Jardine PERSON

Stash Wislocki PERSON

John Schade PERSON

July DATE

Woods Hole LOC

13 CARDINAL

60 miles OUANTITY

Bethel ORG

6,000 CARDINAL

Woods Hole FAC

Sue Natali PERSON

Woods Hole LOC

Alaska GPE

more than 2,100 feet QUANTITY

Vladimir E. Romanovsky PERSON

the University of Alaska, Fairbanks ORG

65 feet QUANTITY

3 degrees Celsius QUANTITY

about 5.5 degrees QUANTITY

Fahrenheit GPE

over decades DATE

one CARDINAL

minus 8 degrees Celsius QUANTITY

Minus 3 LAW

zero CARDINAL

Romanovsky PERSON

the middle of the century DATE

Alaska GPE

Romanovsky PERSON

Bethel GPE

about 4 degrees QUANTITY

the mid-20th century DATE

Bethel LOC

winter DATE

spring DATE

years DATE

Sarah Ludwig PERSON

Woods Hole ORG

Laura Jardine PERSON

Jardine PERSON

Stash Wislocki PERSON

John Schade PERSON

Sarah Ludwig PERSON

Woods Hole ORG

Laura Jardine PERSON

Jardine PERSON

Stash Wislocki PERSON

John Schade Parts PERSON

the 1970s DATE

the summer of 2015 DATE

Bethel GPE

Natali PERSON

one CARDINAL

the century DATE

about 1.5 billion tons QUANTITY

annual DATE

the United States GPE

Alaskan NORP

earlier this year DATE

Alaska GPE

Holmes PERSON

Siberia LOC

Canada GPE

over thousands and thousands CARDINAL

Correction August 23, 2017 EVENT

1.5 billion tons QUANTITY

the century DATE

1.5 billions tons QUANTITY

August 25, 2017 DATE

one CARDINAL

the University of Alaska, Fairbanks ORG

Vladimir E. Romanovsky PERSON

Ramonovsky PERSON

Horowitz PERSON

Muhammadu Buhari PERSON

Monday DATE

Lagos GPE

Ogun GPE

Abuja GPE

last 14 days DATE

Monday DATE

night TIME

14 days DATE

Nigerians NORP

Nigerians NORP

H.E. MUHAMMADU PERSON

BUHARI GPE

THE FEDERAL REPUBLIC OF NIGERIA GPE

19 CARDINAL

THE STATE HOUSE ORG

MONDAY DATE

APRIL DATE

2020 1 DATE

Fellow Nigerians 2. ORG

Sunday, 29th March, 2020 DATE

Lagos GPE

Ogun States GPE

the Federal Capital Territory ORG

an initial period of DATE

fourteen days DATE

Monday, 30th March 2020 DATE

3 CARDINAL

State Governments ORG

4. CARDINAL

5. CARDINAL

COVID-19 ORG

6 CARDINAL

Coronavirus LOC

COVID-19 ORG

the Christian Association of Nigeria ORG

the Nigerian Supreme Council for Islamic Affairs ORG

NSCIA ORG

9 CARDINAL

10 CARDINAL

11 CARDINAL

these 14 days DATE

12 CARDINAL

13 CARDINAL

92% PERCENT

1,500 CARDINAL

7,000 CARDINAL

Healthcare ORG

NCDC ORG

19 CARDINAL

15 CARDINAL

Abuja GPE

today DATE

some 1,000 CARDINAL

16. CARDINAL

State Governments ORG

17 CARDINAL

the coming weeks DATE

18 CARDINAL

19 CARDINAL

Lagos GPE

Ogun States GPE

the Federal Capital Territory ORG

20. CARDINAL

the coming weeks DATE

21 CARDINAL

COVID-19 ORG

22 CARDINAL

Nigeria GPE

23 CARDINAL

30th March 2020 DATE

over 780,000 CARDINAL

24 CARDINAL

Yesterday DATE

COVID-19 PERSON

one million CARDINAL

eight hundred and fifty thousand CARDINAL

two weeks DATE

25 CARDINAL

the last fourteen days alone DATE

over 70,000 CARDINAL

26 CARDINAL

27 CARDINAL

Nigeria GPE

131 CARDINAL

COVID-19 ORG

12 CARDINAL

30th March 2020 DATE

two CARDINAL

28 CARDINAL

This morning TIME

Nigeria GPE

323 CARDINAL

twenty CARDINAL

ten CARDINAL

Lagos State GPE

54% PERCENT

Nigeria GPE

FCT ORG

two CARDINAL

71% PERCENT

Nigeria GPE

29 CARDINAL

two CARDINAL

30 CARDINAL

Lagos GPE

FCT ORG

31 CARDINAL

States GPE

32 CARDINAL

The National Centre for Disease Control ORG

33 CARDINAL

Nigerians NORP

34. CARDINAL

the Quarantine Order ORG

COVID-19 ORG

35 CARDINAL

COVID-19 ORG

Today DATE

37 CARDINAL

2 CARDINAL

First ORDINAL

Nigerians NORP

second ORDINAL

38. CARDINAL

Report ORG

the Presidential Task Force ORG

Lagos GPE

Ogun States GPE

FCT ORG

another 14 days DATE

11:59 pm TIME

Monday, 13th of April, 2020 DATE

39 CARDINAL

Mosques FAC

Makkah ORG

Madina GPE

St. Peter's Square GPE

Notre Dame FAC

Paris GPE

Easter Mass FAC

less than 10 CARDINAL

India GPE

Italy GPE

France GPE

```
40 CARDINAL
```

41 CARDINAL

42 CARDINAL

45 CARDINAL

daily DATE

46 CARDINAL

47. CARDINAL

the past two weeks DATE

48 CARDINAL

2.6 million CARDINAL

3.6 million CARDINAL

the next two weeks DATE

an additional one million CARDINAL

the end of this week DATE

49 DATE

50 CARDINAL

Fellow Nigerians PERSON

51 CARDINAL

State Governors ORG

the Federal Government ORG

52 CARDINAL

the coming weeks DATE

the Federal Government ORG

the Presidential Task Force ORG

54 CARDINAL

55 CARDINAL

the Ministers of Industry, Trade and Investment, Communication ORG

Digital Economy, Science and Technology ORG

Transportation ORG

Interior ORG

Health, Works and Housing, Labour and Employment and Education ORG

Nigerian NORP

COVID-19 ORG

56 CARDINAL

the Presidential Economic Advisory Council and Economic Sustainability

Committee ORG

National Food Security Council ORG

the Presidential Task Force ORG

COVID-19 ORG

2020 DATE

the Presidential Task Force ORG

COVID-19 ORG

Fellow Nigerians PERSON

God PERSON

the Federal Republic of Nigeria GPE

COVID-19 PERSON

The Centers for Disease Control ORG

COVID-19 ORG

99.99 percent PERCENT

Plexiglas PERSON

American NORP

one CARDINAL

U.S. GPE

the Centers for Disease Control ORG

Anthony Fauci PERSON

Chicago GPE

YouTube ORG

Twitter PERSON

more than 26,000 CARDINAL

YouTube ORG

Pennsylvania GPE

Tom Wolf PERSON

Faking 'Science' WORK\_OF\_ART

Achieve Political Goals ORG

's NORP

The Centre for Evidence-Based Medicine ORG

Oxford University ORG

six CARDINAL

ILI LOC

Oxford NORP

two decades DATE

more than 1,600 CARDINAL

3 percent PERCENT

the New England Journal of Medicine ORG

the National Center for Biotechnology Information ORG

Cambridge University Press ORG

Oxford Clinical Infectious Diseases ORG

Influenza Journal ORG

two CARDINAL

two CARDINAL

**Graphs PERSON** 

Austria GPE

one CARDINAL

first ORDINAL

about 10 days DATE

four CARDINAL

Austria GPE

Germany GPE

French NORP

around 1,000 percent PERCENT

daily DATE

one CARDINAL

Spain GPE

French NORP

Spain GPE

zero CARDINAL

Europe LOC

Spain GPE

around 1,500 percent PERCENT

three months DATE

the United Kingdom GPE

around 1,500 percent PERCENT

Europe LOC

Belgium GPE

British NORP

Italy GPE

Ireland GPE

Portugal GPE

Israel GPE

the United States GPE

Americans NORP

Germans NORP

Europe LOC

U.S. GPE

California GPE

June DATE

more than 300 percent PERCENT

four months later DATE

Hawaii GPE

one CARDINAL

U.S. GPE

almost 1,000 percent PERCENT

Texas GPE

Georgia GPE

The Food and Drug Administration ORG

CDC ORG

Sweden GPE

Georgia GPE

first ORDINAL

COVID-19 ORG

Nobel WORK OF ART

COVID ORG

Martin Kulldorff PERSON

Harvard Medical School ORG

COVID ORG

300 years DATE

the Age of Enlightenment EVENT

AdvertisementsOne day DATE

the February 3 EVENT

Iowa GPE

first ORDINAL

fifty CARDINAL

American NORP

Twitter PERSON

a confusing night TIME

Iowa GPE

One CARDINAL

first ORDINAL

Pete Buttigieg PERSON

Iowa GPE

at least one CARDINAL

Pete Buttigieg's PERSON

Atlantic LOC

February 3 2020 DATE

BURLINGTON GPE

Iowa GPE

Democratic NORP

Iowans NORP

hours TIME

the end of the night DATE

Joe Biden PERSON

Elizabeth Warren PERSON

Bernie Sanders PERSON

Sandy Dockendorff PERSON

tonight TIME

62-year-old DATE

Danville ORG

late January 2020 DATE

Democrats NORP

Iowa GPE

first ORDINAL

Dockendorff PERSON

Iowa GPE

Eastern ORG

the Iowa State Democratic Party ORG

Twitter PRODUCT

Iowa GPE

**#1 CARDINAL** 

Robby Mook PERSON

Hillary Clinton's PERSON

2016 DATE

Iowa GPE

Mook PERSON

Monday night TIME

Iowa GPE

Shadow PRODUCT

Mook PERSON

January 29 2020 DATE

the Des Moines Register ORG

Mook PERSON

Iowa GPE

fall 2018 DATE

Harvard ORG

night TIME

Robby Mook PERSON

Matt Rhodes PERSON

Mook PERSON

2016 DATE

Hillary Clinton PERSON

Rhodes GPE

Mitt Romney's PERSON

2012 DATE

2018 DATE

days DATE

Russian NORP

Clinton PERSON

DNC ORG

the Democratic Congressional Campaign Committee ORG

2016 DATE

Mook PERSON

February 4 2020 DATE

Clinton PERSON

Iowa GPE

Nevada GPE

February 3 2020 DATE

Iowa GPE

Iowa GPE

Democrats NORP

Hillary Clinton PERSON

2016 DATE

the Nevada Democratic Party ORG

February 4 2020 DATE

Feb. 22 DATE

2020 CARDINAL

Jimmy Hickey PERSON

the Des Moines Register ORG

Tuesday DATE

Gerard Niemira PERSON

Krista Davis PERSON

Clinton PERSON

2016 DATE

Shadow PRODUCT

Iowa GPE

the Democratic National Committee ORG

DNC ORG

Bernie Sanders ORG

2016 DATE

Bernie Sanders PERSON

2016 DATE

Iowa GPE

2020 DATE

February 3 2020 DATE

Associated Press ORG

Democrats NORP

DNC ORG

Bernie Sanders PERSON

Hillary Clinton PERSON

2016 DATE

Iowa GPE

Nevada GPE

Democratic NORP

Feb. 22 DATE

three CARDINAL

two CARDINAL

Iowa GPE

Sanders ORG

2016 CARDINAL

December 2017 DATE

Des Moines Register ORG

2020 DATE

Iowans NORP

2020 DATE

Iowa GPE

first ORDINAL

Democratic NORP

decades DATE

2020 DATE

A Democratic National Committee ORG

the Unity Reform Commission ORG

December 2017 DATE

Iowa GPE

five decades ago DATE

DNC ORG

Tom Perez PERSON

Perez PERSON

2016 DATE

Jeff Weaver PERSON

2016 DATE

Jeff Weaver PERSON

2016 CARDINAL

Martin O'Malley PERSON

Iowa GPE

0.6 percent PERCENT

Iowa GPE

at least two CARDINAL

DNC ORG

2016 CARDINAL

Bernie Sanders PERSON

Hillary Clinton's PERSON

Sanders ORG

Weaver ORG

DNC ORG

2020 DATE

Iowa GPE

Iowa GPE

**#3 CARDINAL** 

Iowa GPE

Democratic Party ORG

50 percent PERCENT

5PM CARDINAL

February 4 2020 DATE

the Democratic Party ORG

the morning TIME

Iowa GPE

25% PERCENT

the Iowa Democratic Party ORG

The Iowa Democratic Party ORG

three CARDINAL

1:14 PM EST TIME

February 4 2020 DATE

NPR ORG

Tuesday DATE

afternoon TIME

February 4 2020 DATE

Democratic Party ORG

5 p.m. ET TIME

February 4 2020 DATE

Tuesday DATE

February 4 2020 DATE

Troy Price ORG

NPR ORG

Christine Brown PERSON

late January 2020 DATE

the Asian/Pacific Islander EVENT

the Iowa Democratic Party ORG

first ORDINAL

the evening TIME

January 31 2020 DATE

**Brown PERSON** 

NPR ORG

Rachel Martin PERSON

**Brown PERSON** 

February 3 2020 DATE

Iowa GPE

Elesha Gayman PERSON

Scott County GPE

PIN ORG

Des Moines GPE

Tom Courtney PERSON

Courtney PERSON

several hours TIME

NPR ORG

Nevada Democratic Party ORG

February 22 2020 DATE

NV Dems ORG

Iowa GPE

last night TIME

Nevada GPE

February 22nd DATE

2020 CARDINAL

Iowa GPE

2016 DATE

Iowa GPE

Iowa GPE

Twitter PRODUCT

2016 DATE

the day DATE

Iowa GPE

Shadow PRODUCT

Iowa GPE

Democratic Party ORG

Iowa GPE

2020 DATE

day DATE

Invalid ORG

Yesterday DATE

UK GPE

Fathers 4 Justice ORG

Hallmark GPE

ComRes ORG

more than 1000 CARDINAL

37 per cent MONEY

UK GPE

2.3 million CARDINAL

18 DATE

one CARDINAL

five CARDINAL

20 per cent MONEY

22 per cent MONEY

One third 32 CARDINAL

Matt O'Connor PERSON

Fathers 4 Justice ORG

Telegraph ORG

daily DATE

millions CARDINAL

Women ORG

ComRes ORG

nearly one CARDINAL

six CARDINAL

17 per cent MONEY

quarter 27 DATE

2001 DATE

750,000 CARDINAL

roughly one CARDINAL

four CARDINAL

10 CARDINAL

UK GPE

today DATE

first ORDINAL

a month ago DATE

Today DATE

Nicola ORG

CBE ORG

Chris Snowden PERSON

Julia Goodfellow PERSON

first ORDINAL

around 100 years DATE

UUK ORG

the last 5 years DATE

David Willetts PERSON

**Greg Clark PERSON** 

OECD ORG

Graduate Labour Market ORG

2007 DATE

87.5% PERCENT

UK GPE

HE GPE

4 CARDINAL

10 CARDINAL

Unfinished PERSON

the coming months DATE

3 CARDINAL

firstly ORDINAL

2020 DATE

2009 DATE

secondly ORDINAL

first ORDINAL

2012 DATE

HEPI ORG

BBC/ComRes ORG

only around half CARDINAL

the Small Business, Enterprise and Employment Act ORG

March DATE

Department for Business, Innovation and Skills BIS and ORG

Department for Education DfE ORG

Department for Work and Pensions DWP ORG

National Student Survey ORG

US GPE

the Competition and Markets Authority ORG

Higher Education Policy Institute Hepi ORG

as many as three-quarters CARDINAL

the last 3 years DATE

1992 DATE

Between 2006 and 2015 DATE

around 55% PERCENT

around 45% PERCENT

22,000 MONEY

Last year DATE

CBI /Pearson Education ORG

Skills ORG

```
47% PERCENT
```

Excellence Framework PERSON

Parliament ORG

David Willetts PERSON

the Times Higher Education ORG

English LANGUAGE

the Research Excellence Framework ORG

the Teaching Excellence Framework ORG

TEF ORG

TEF ORG

TEF ORG

REF ORG

the Higher Education Funding Council for England ORG

the Quality Assurance Agency ORG

2010 DATE

the Department of Business, Innovation ORG

Skills PRODUCT

the autumn DATE

One CARDINAL

TEF ORG

UK GPE

Last week DATE

2015 DATE

the Higher Education Achievement Report ORG

2:1 degrees QUANTITY

2:1s CARDINAL

UK GPE

300% PERCENT

the 1990s DATE

Over 70% PERCENT

First ORDINAL

2:1 CARDINAL

7 CARDINAL

the past 5 years DATE

just 47% PERCENT

the mid-1990s DATE

2013 DATE

14 CARDINAL

over 50% PERCENT

second ORDINAL

The Higher Education Academy ORG

nearly half CARDINAL

The Teaching Excellence Framework PERSON

the Teaching Excellence Framework ORG

First ORDINAL

2:1 CARDINAL

2:2 CARDINAL

Third ORDINAL

GPA ORG

13 CARDINAL

2:1 CARDINAL

the Teaching Excellence Framework ORG

Student Academic Experience Survey ORG

30.5 hours TIME

14 hours TIME

weekly DATE

Mass Communications and Documentation ORG

as little as 22 hours QUANTITY

GPA ORG

Robert Burgess PERSON

the Grade Point Average Advisory Group ORG

UK GPE

the end of the decade DATE

GPA ORG

Higher Education Academy ORG

2012 DATE

this academic year DATE

2020 CARDINAL

2009 DATE

Les Ebdon PERSON

the Office of Fair Access ORG

Degree Apprenticeships PERSON

2-year DATE

the National Strategy for Access and Student Success ORG

their college years DATE

years DATE

November 8, 2016 DATE

New Hampshire's GPE

U.S. Senate Seat ORG

New Hampshire's GPE

four CARDINAL

New Hampshire GPE

one CARDINAL

fifteen CARDINAL

same-day DATE

election day DATE

same-day DATE

U.S. GPE

New Hampshire GPE

Massachusetts GPE

Vermont GPE

Democrat NORP

New Hampshire GPE

New Hampshire's GPE

the Granite State ORG

the New Hampshire House of Representatives ORG

November 2016 DATE

6,540 CARDINAL

New Hampshire GPE

State ORG

New Hampshire GPE

New Hampshire GPE

60 days DATE

New Hampshire GPE

6,540 CARDINAL

New Hampshire GPE

no later than January 7, 2017 DATE

August 30, 2017 DATE

nearly ten months DATE

only 1,014 CARDINAL

6,540 CARDINAL

New Hampshire GPE

5,526 CARDINAL

New Hampshire GPE

5,526 CARDINAL

only 213 CARDINAL

New Hampshire GPE

5,313 CARDINAL

New Hampshire GPE

New Hampshire GPE

New Hampshire GPE

State ORG

5,513 CARDINAL

two CARDINAL

Republican NORP

U.S. GPE

Kelly Ayotte PERSON

Maggie Hassan PERSON

Hassan PERSON

1,017 CARDINAL

5,313 CARDINAL

59.2 percent PERCENT

Hassan PERSON

Democrat NORP

New Hampshire's GPE

four CARDINAL

Hillary Clinton PERSON

Clinton PERSON

New Hampshire GPE

only 2,732 CARDINAL

74.8 percent PERCENT

5,513 CARDINAL

Clinton PERSON

New Hampshire GPE

three CARDINAL

four CARDINAL

Senate ORG

52 CARDINAL

Republican NORP

Democrat NORP

Senate ORG

53-47 TIME

Obamacare ORG

Kris W. Kobach PERSON

State ORG

Kansas GPE

Arizona GPE

10 CARDINAL

Obama's ORG

2012 DATE

2017 DATE

Trump PERSON

the Presidential Commission on Election Integrity ORG

Kansas GPE

Miami Beach GPE

Bruce A. Mowry PERSON

Miami Beach GPE

Biscayne Bay PERSON

Miami Beach GPE

at least \$400 million MONEY

2018 DATE

Miami-Dade County GPE

billions CARDINAL

Harvey Ruvin PERSON

Miami-Dade County GPE

Florida GPE

a half-century DATE

Republican NORP

Democrats NORP

James C. Cason PERSON

Republican NORP

Coral Gables ORG

hundreds CARDINAL

coming years DATE

Cason PERSON

South Florida LOC

Elisa Granato PERSON

first ORDINAL

Oxford ORG

first ORDINAL

Europe LOC

two days DATE

Elisa PERSON

few hours TIME

Elisa Granato PERSON

two CARDINAL

Four CARDINAL

four CARDINAL

Elisa Granato PERSON

BBC ORG

Elisa PERSON

four CARDINAL

Nearly 99 CARDINAL

Indonesian NORP

780 CARDINAL

98.9 per cent MONEY

just 4.1 CARDINAL

Covid-19 ORG

One CARDINAL

Anglia Ruskin University ORG

Cambridge GPE

European NORP

The National Institute for Health and Care Excellence ORG

as early as next week DATE

Vitamin D PERSON

the University of East Anglia ORG

May DATE

million CARDINAL

100,000 CARDINAL

Asian NORP

Indonesian NORP

five CARDINAL

Prabowo Raharusuna PERSON

April DATE

10 CARDINAL

One CARDINAL

five CARDINAL

British NORP

one CARDINAL

six CARDINAL

1billion CARDINAL

months DATE

the Scientific Advisory Committee on Nutrition ORG

VITAMIN ORG

Asian NORP

NHS ORG

African NORP

African-Caribbean NORP

Asian NORP

daily DATE

the year DATE

Vitamin D PERSON

BAME PERSON

UK GPE

William Henley PERSON

University of Exeter ORG

MailOnline ORG

COVID-19 ORG

UK GPE

European NORP

winter months DATE

African NORP

African-Caribbean NORP

Asian NORP

Boris Johnson PERSON

Tory PERSON

two-metre QUANTITY

Culture NORP

Oliver Dowden PERSON

50 CARDINAL

Friday DATE

More than 1,600 CARDINAL

months DATE

2.2million CARDINAL

England GPE

100-day DATE

Vitamin D PERSON

as little as 3p PERCENT

Public Health England ORG

April DATE

Adrian Martineau PERSON

Queen Mary University ORG

London GPE

Vitamin PERSON

Guardian ORG

last night TIME

the University of Surrey ORG

the British Medical Journal ORG

last month DATE

Lee Smith PERSON

Anglia Ruskin's PERSON

75 per cent MONEY

D. NORP

Public Health England ORG

NHS ORG

The Institute of Medicine ORG

US GPE

100 CARDINAL

Dan PERSON

the Wisconsin Elections Commission ORG

this November DATE

February 3, 2021 Perspective DATE

Dan O'Donnell Calling PERSON

November DATE

these days DATE

One CARDINAL

U.S. GPE

Capitol FAC

Wisconsin GPE

The Wisconsin Elections Commission ORG

WEC ORG

234,000 CARDINAL

Kanye West PERSON

the Green Party ORG

Joe Biden PERSON

Milwaukee GPE

Dane County GPE

Madison PERSON

Biden PERSON

WEC ORG

Election Data Report PRODUCT

November DATE

nearly zero CARDINAL

Just 4,270 CARDINAL

1,969,274 CARDINAL

0.2% PERCENT

November DATE

the spring 2016 DATE

four years ago DATE

2.5% PERCENT

first ORDINAL

0.07% PERCENT

1,434 CARDINAL

0.05% PERCENT

1,045 CARDINAL

0.002% PERCENT

iust 42 CARDINAL

1.8% PERCENT

just seven months earlier DATE

Wisconsin GPE

spring DATE

1,101,474 CARDINAL

78% PERCENT

November DATE

78% PERCENT

hundreds of thousands CARDINAL

Wisconsinites NORP

April 2016 DATE

2.5% PERCENT

1.4% PERCENT

November DATE

2020 DATE

April 2020 DATE

1.8% PERCENT

0.2% PERCENT

November 2018 DATE

Milwaukee GPE

Tony Evers PERSON

Scott Walker PERSON

0.2% PERCENT

```
two CARDINAL
```

Republicans NORP

Donald Trump PERSON

the Wisconsin Elections Commission ORG

2016 DATE

WEC ORG

2018 DATE

2020 CARDINAL

two CARDINAL

Democrat NORP

Republican NORP

the 23 June DATE

Britain GPE

Britain GPE

Europe LOC

Britain GPE

Europe LOC

Britain GPE

Europe LOC

Almost half CARDINAL

Europe LOC

Centre for Economics and Business Research ORG

up to four million CARDINAL

British NORP

Europe LOC

Britain GPE

26.5bn MONEY

EU ORG

the last few years DATE

BMW ORG

7,000 CARDINAL

Oxford GPE

Birmingham GPE

Swindon ORG

UK GPE

Europe LOC

88 per cent MONEY

EU ORG

the Bank of England ORG

Europe LOC

Europe LOC

Europe LOC

Erasmus PRODUCT

Britain GPE

Europe LOC

a year DATE

Britain GPE

Europe LOC

EU ORG

French NORP

German NORP

10 per cent MONEY

Europe LOC

Europe LOC

One CARDINAL

London GPE

2007 DATE

Italy GPE

Britain GPE

the European Arrest Warrant ORG

EU ORG

over 5,000 CARDINAL

EU ORG

2010 DATE

European NORP

Britain GPE

EU ORG

the United Nations ORG

Nato ORG

EU ORG

Russia GPE

Ukraine GPE

the Horn of Africa LOC

British NORP

1bn ORDINAL

West Africa GPE

Europe LOC

first ORDINAL

EU ORG

Europe LOC

Norway GPE

Switzerland GPE

Europe LOC

Russia GPE

Europe LOC

EU ORG

EU ORG

thousands CARDINAL

Europe LOC

Europe LOC

Britain GPE

British NORP

Outers PRODUCT

Europe LOC

EU ORG

790,000 CARDINAL

UK GPE

2030 DATE

58bn MONEY

Government ORG

Britain GPE

Europe LOC

Europe LOC

Europe LOC

Britain GPE

Europe LOC

Europe LOC

British NORP

Straw ORG

Britain Stronger ORG

Europe LOC

Biden PERSON

GEORGIA GPE

Donald Trump PERSON

2015 DATE

Muslims NORP

the United States GPE

Europe LOC

ISIS ORG

weekly DATE

hundreds CARDINAL

85 CARDINAL

Paris GPE

German NORP

Christian NORP

Islam ORG

Islam ORG

the United States GPE

Georgia GPE

Muslim NORP

Jason Spencer PERSON

Wednesday DATE

Spencer PERSON

Islamic NORP

Spencer PERSON

Islamic NORP

the State of GPE

Georgia GPE

Spencer PERSON

WSB-TV ORG

Atlanta GPE

the First Amendment LAW

Islam ORG

Christians NORP

Bible WORK\_OF\_ART

Quran GPE

Quran GPE

Allah GPE

the Last Day DATE

His Messenger PERSON

Islam ORG

Jews NORP

Christians NORP

Jizya LOC

Allah GPE

the Last Day DATE

His Messenger PERSON

Islam ORG

Jews NORP

Christians NORP

Jizya LOC

Quran GPE

Unbelievers PERSON

Allah ORG

Liberals NORP

Christian NORP

Spencer PERSON

Liberals ORG

the United States GPE

Georgia GPE

Islam ORG

France GPE

Belgium GPE

Italy GPE

Spain GPE

Switzerland GPE

American NORP

Trump PERSON

American NORP

Control of Infectious Diseases ORG

41 CARDINAL

the House of Representatives ORG

Nigerians NORP

the House of Representatives ORG

the Ouarantine Act LAW

the Control of Infectious Diseases Bill ORG

second ORDINAL

last week DATE

82 CARDINAL

the House of Representatives ORG

Gbajabiamila ORG

the Nigeria Centre for Disease Control ORG

NCDC ORG

NCDC ORG

Nigeria GPE

Tuesday DATE

NCDC ORG

Bill PERSON

NCDC ORG

Section 10 LAW

3 CARDINAL

Nigeria GPE

Center For Liberty ORG

Civil Society Legislative Advocacy Centre ORG

Amnesty International ORG

YIAGA Africa GPE

Human and Environmental Development Agenda ORG

Global Rights and African Centre for Media and Information Literacy ORG

are International Press Centre ORG

Lawyers Alert PERSON

Women Advocates Research and Documentation Centre ORG

Enough, ORG

Community Life Project, Centre for Information Technology and

Development, Centre for Impact Advocacy ORG

Concerned Nigerians ORG

Lex Community NG ORG

Education ORG

Vaccine PERSON

Dinidari Foundation ORG

Albino Foundation ORG

The Community of PWDs ORG

Nigeria GPE

Dorothy Njemanze Foundation ORG

Tap Nitiative ORG

Raising New Voices Initiative ORG

Haly Hope Foundation ORG

Youths Concerns Development Foundation ORG

PITCH Nigeria GPE

House of Justice ORG

Molluma Medico-Legal Centre PERSON

Albino Foundation ORG

The Community of PWDs ORG

Nigeria GPE

Alliances GPE

Africa LOC

Motion, Persons with Disabilities Action Network ORG

Tuesday DATE

Gbajabiamila PERSON

Nigerians NORP

Nigerian NORP

House ORG

Robin Williams PERSON

Last week DATE

Home Office ORG

a decade DATE

Nearly 9 per cent MONEY

16 to CARDINAL

24-year-olds DATE

the past 12 months DATE

Britain GPE

Europe LOC

Global sea LOC

the end of the century DATE

another 2 feet QUANTITY

CNN ORG

Monday DATE

the National Academy of Sciences ORG

University of Colorado-Boulder ORG

Steve Nerem PERSON

1993 DATE

between 1992 and 2014 DATE

7 centimeters QUANTITY

2.8 inches QUANTITY

25 years DATE

about 3 CARDINAL

0.1 inches QUANTITY

Read More WORK\_OF\_ART

Leon Young PERSON

the Wisconsin Legislature ORG

Earlier this week DATE

Young PERSON

Milwaukee GPE

the first three months of the year DATE

75 percent PERCENT

12,000 MONEY

49,943 MONEY

2011-12 DATE

only 34 days DATE

122 session days DATE

Young PERSON

2012 DATE

Young PERSON

seven CARDINAL

seven days DATE

a whole year DATE

the National Conference of State Legislatures ORG

Wisconsin GPE

10 CARDINAL

the Milwaukee County Board ORG

Republicans NORP

Legislature ORG

Alberta Darling PERSON

Joe Sanfelippo PERSON

15,000 MONEY

51,000 MONEY

\$6.5 million MONEY

annual DATE

Legislature ORG

Milwaukee County Board ORG

Young PERSON

recent days DATE

Democracy Campaign The Wisconsin Democracy Campaign ORG

the holidays DATE

Madison PERSON

the Joyce Foundation ORG

16 years DATE

Mike McCabe PERSON

232,000 MONEY

Joyce PERSON

The Democracy Campaign's ORG

annual DATE

430,000 MONEY

McCabe ORG

roughly \$130,000 MONEY

Capitol FAC

McCabe ORG

McCabe ORG

the Democracy Campaign ORG

McCabe ORG

Capitol ORG

Michigan GPE

Michigan GPE

Alcona County GPE

73rd ORDINAL

1998 DATE

the Associated Press ORG

Alcona GPE

Alpena GPE

Oscoda ORG

TB ORG

the Michigan Department of Natural Resources ORG

Michigan GPE

Fox-TV ORG

Mobile GPE

Ala. GPE

State ORG

Nancy Barr PERSON

the Associated Press ORG

5 Million Muslim Children MONEY

Yemen GPE

Hunger, Wars, COVID-19 & Poverty ORG

2020 DATE

the past two years DATE

2018-2020 DATE

85,000 CARDINAL

Yemen GPE

the year 2018 DATE

one CARDINAL

UK GPE

```
5 million CARDINAL
Yemen GPE
2018 DATE
2020 DATE
UK GPE
more than 5 million CARDINAL
Yemen GPE
2020 DATE
COVID-19 PERSON
Muslims NORP
Madina GPE
COVID-19 ORG
Helle Thorning-Schmidt PERSON
Save the Children International ORG
Yemen GPE
cholera & off ORG
COVID-19" Save the Children Organisation ORG
2018 DATE
four million CARDINAL
Save The Children ORG
Wednesday DATE
another million CARDINAL
Hodeida NORP
2 years DATE
Yemen GPE
one CARDINAL
Yemen GPE
one CARDINAL
Yemen GPE
Thorning-Schmidt PERSON
Telegraph FAC
UK GPE
2018 DATE
5.2 million CARDINAL
Yemen GPE
Britain GPE
NGO ORG
the year 2018 DATE
more than 10 CARDINAL
Yemen GPE
Yemen GPE
up to 85,000 CARDINAL
Guardian ORG
More than five million CARDINAL
Yemen GPE
Unicef Usa PERSON
Yemen GPE
Yemen GPE
Children Are on the Edge of Starvation WORK OF ART
Muslim NORP
```

Arab NORP

Saudi Arabia's GPE

Muslim NORP

Holy Prophet Muhammad PERSON

over 20 hours TIME

Friday DATE

Breonna Taylor PERSON

Taylor PERSON

today DATE

Taylor PERSON

Taylor PERSON

months DATE

months ago DATE

Taylor PERSON

the Daily Mail ORG

Taylor PERSON

Taylor PERSON

Brett Hankison PERSON

Taylor PERSON

Kenneth Walker PERSON

One CARDINAL

Taylor PERSON

Officers PRODUCT

Taylor PERSON

the night TIME

almost a minute TIME

10 or 15 seconds TIME

Louisville GPE

Kentucky GPE

Shawn Hoover PERSON

Taylor PERSON

third ORDINAL

45 seconds TIME

a minute TIME

Hoover PERSON

Walker ORG

One CARDINAL

Taylor PERSON

4 CARDINAL

Breonna Taylor's PERSON

Taylor PERSON

Louisville GPE

March 13 DATE

26-year-old DATE

five CARDINAL

Cameron ORG

two CARDINAL

Taylor PERSON

Taylor PERSON

first ORDINAL

Brett Hankison PERSON

three CARDINAL

American NORP

George Floyd PERSON

Taylor PERSON

Jason Whitlock PERSON

Taylor PERSON

BILL GATES EXPLAINS THAT THE COVID VACCINE ORG

EXPERIMENTAL ORG

the Mother Jones Daily ORG

Newt Gingrich PERSON

one CARDINAL

Donald Trump's PERSON

Cabinet ORG

Republican NORP

Trump ORG

Hillary Clinton PERSON

Gingrich PERSON

two years ago DATE

Trump ORG

four years ago DATE

Electoral College ORG

Election Day 2012 DATE

Mitt Romney PERSON

the Electoral College ORG

Barack Obama PERSON

Trump ORG

April DATE

Ben Carson PERSON

the Electoral College ORG

Gingrich PERSON

Trumpsters ORG

Trump ORG

2014 DATE

Gingrich PERSON

National Popular Vote ORG

11 CARDINAL

one CARDINAL

an additional 12 CARDINAL

Americans NORP

two years ago DATE

Gingrich PERSON

Gingrich PERSON

Americans NORP

Gingrich PERSON

314 million CARDINAL

Earth LOC

Gingrich PERSON

America GPE

Clinton PERSON

Trump ORG

60 million CARDINAL

59.8 million CARDINAL

one-fifth CARDINAL

Americans NORP

Gingrich PERSON

Trump ORG

one CARDINAL

the White House FAC

Trump ORG

2012 DATE

The United States GPE

RT-PCR ORG

Torsten Engelbrecht PERSON

Konstantin Demeter Lockdowns PERSON

PCR ORG

COVID-19 ORG

March 16, 2020 DATE

WHO ORG

Tedros Adhanom Ghebreyesus PERSON

Reuters ORG

BBC ORG

3 CARDINAL

May DATE

German NORP

PCR ORG

Walter Lippmann PERSON

two CARDINAL

Pulitzer Prize WORK OF ART

the 20th century DATE

Kary Mullis PERSON

the Polymerase Chain Reaction PCR ORG

the Nobel prize WORK OF ART

1993 DATE

Mullis PERSON

last year DATE

the age of 74 DATE

PCR ORG

PCR ORG

millions and billions MONEY

PCR ORG

Gina Kolata PERSON

2007 DATE

New York Times ORG

Faith WORK\_OF\_ART

PCR ORG

COVID-19 PRODUCT

Australian NORP

Sanjaya Senanayake PERSON

ABC TV ORG

decades DATE

COVID-19 PERSON

Jessica C. Watson PERSON

Bristol University ORG

The British Medical Journal ORG

COVID-19 ORG

COVID-19 PERSON

Watson PERSON

PCR ORG

PCR ORG

PCR ORG

COVID-19 ORG

Thomas Löscher PERSON

the Department of Infection and Tropical Medicine ORG

the University of Munich ORG

the Federal Association of German Internists ORG

COVID-19 ORG

COVID-19 PERSON

Watson PERSON

Watson PERSON

Watson PERSON

COVID-19 PERSON

COVID-19 ORG

June 2nd DATE

later this week DATE

RNA ORG

first ORDINAL

RNA ORG

PCR ORG

White/Fenner ORG

1986 DATE

9 CARDINAL

Luc Montagnier PERSON

Dominic Dwyer PERSON

Nobel WORK OF ART

Marie Curie PERSON

100 CARDINAL

1898 DATE

RNA ORG

PCR ORG

PCR ORG

RNA ORG

RNA ORG

NB ORG

Leo L. M. Poon PERSON

Malik Peiris PERSON

Nature Medicine WORK OF ART

March 2020 DATE

Malik Peiris PERSON

May 12 DATE

2020 DATE

2 CARDINAL

Myung-Guk Han et al ORG

Korea GPE

COVID-19 ORG

Osong Public Health and Research Perspectives ORG

February 2020 DATE

Myung-Guk Han Date PERSON

May 6, 2020 Answer: DATE

3 CARDINAL

Wan Beom Park PERSON

the First Patient ORG

Korea GPE

Journal of Korean Medical Science ORG

February 24 DATE

2020 DATE

Wan Beom Park Date: PERSON March 19, 2020 Answer: DATE

4 CARDINAL

Na Zhu et al. PERSON

A Novel Coronavirus from Patients WORK\_OF\_ART

China GPE 2019 DATE

New England Journal of Medicine ORG

February 20, 2020 DATE

Wenjie Tan Date: PERSON

March 18, 2020 DATE

four CARDINAL

early 2020 DATE

Charles Calisher PERSON

2001 DATE

Calisher PERSON

Dr Calisher PERSON

one CARDINAL

RNA ORG

PCR ORG

RNA ORG

COVID-19 ORG

one CARDINAL

four CARDINAL

Koch PERSON

Amory Devereux ORG

Rosemary Frei PERSON

OffGuardian NORP

One CARDINAL

Nature GPE

May 7 DATE

third ORDINAL

German NORP

Alexander S. Kekulé University of Halle ORG

Hartmut Hengel PERSON

Ralf Bartenschlager German PERSON

Society for Virology ORG

Thomas Löscher PERSON

Ulrich Dirnagl PERSON

Bornkamm PERSON

RNA ORG

Nobel WORK OF ART

Barbara McClintock PERSON

her Nobel Lecture back WORK OF ART

1983 DATE

Charité NORP

Christian Drosten ORG

Germany GPE

COVID-19 ORG

German NORP

PCR ORG

first ORDINAL

WHO ORG

June 18, 2020 DATE

months DATE

Berlin GPE

Viviane Fischer PERSON

Charité NORP

Charité NORP

Charité NORP

Corman NORP

RNA ORG

the MagNA Pure 96 PRODUCT

Roche ORG

Penzberg GPE

Germany GPE

RNA ORG

QIAGEN ORG

Hilden GPE

Germany GPE

RNA ORG

January 23, 2020 DATE

two CARDINAL

PCR ORG

February DATE

China GPE

Guangdong province GPE

COVID-19 ORG

A month later DATE

the Journal of Medical Virology ORG

29 CARDINAL

610 CARDINAL

Wuhan GPE

3 CARDINAL

```
6 CARDINAL
third ORDINAL
Singapore GPE
almost daily DATE
18 CARDINAL
up to five CARDINAL
one CARDINAL
Wang Chen PERSON
the Chinese Academy of Medical Sciences ORG
February DATE
PCR ORG
only 30 to 50 CARDINAL
Sin Hang Lee PERSON
the Milford Molecular Diagnostics Laboratory ORG
WHO ORG
Anthony S. Fauci PERSON
March 22, 2020 DATE
RNA ORG
PCR ORG
PPV ORG
PPV ORG
two CARDINAL
95% PERCENT
5 CARDINAL
100 CARDINAL
PPV ORG
June 12 2020 DATE
Deutsches Ärzteblatt ORG
PPV ORG
three CARDINAL
second ORDINAL
Jessica Watson PERSON
95% PERCENT
Deutsches Ärzteblatt PERSON
PPV ORG
one CARDINAL
three CARDINAL
3% PERCENT
PPV ORG
only 30 percent PERCENT
70 percent PERCENT
Ärzteblatt PERSON
second ORDINAL
20 percent PERCENT
PPV ORG
78 percent PERCENT
22 percent PERCENT
9 million CARDINAL
almost 2 million CARDINAL
```

CDC ORG

FDA ORG

CDC ORG

2019 DATE

2019 DATE

March 30, 2020 DATE

RNA ORG

2019 DATE

FDA ORG

PCR ORG

Altona Diagnostics ORG

the LightMix Modular Assays ORG

TIB Molbiol ORG

Roche ORG

PCR ORG

Roche ORG

aPCR ORG

millions and millions CARDINAL

CDC ORG

WHO ORG

FDA ORG

RKI ORG

Jon Rappoport PERSON

RNA ORG

PCR ORG

a few hundred or even CARDINAL

thousand CARDINAL

PCR ORG

29 CARDINAL

86 DATE

199 CARDINAL

272 CARDINAL

293 CARDINAL

un ORG

Viviane Fischer PERSON

Charité NORP

Charité NORP

Charité NORP

RT-PCR ORG

Charité NORP

the Institut Pasteur ORG

Asian NORP

Asian NORP

the beginning of April DATE

WHO ORG

Asian NORP

Roche ORG

TIB Molbiol ORG

PCR ORG

35 DATE

45 CARDINAL

```
Cq NORP
Cq PERSON
higher than 40 CARDINAL
MIOE ORG
MIOE ORG
Minimum Information for Publication of Quantitative Real-Time PCR
Experiments WORK OF ART
Real-Time PCR ORG
PCR ORG
qPCR NORP
Kary Mullis PERSON
more than 40 CARDINAL
PCR ORG
MIQE ORG
Stephen A. Bustin PERSON
PCR ORG
A-Z of Quantitative PCR WORK OF ART
qPCR NORP
Bustin NORP
Cq PERSON
the 20s to 30s DATE
Cq PERSON
35 DATE
Cq GPE
PCR ORG
RNA ORG
RNA ORG
Reverse Transcriptase PERSON
qPCR PRODUCT
Jessica Schwaber PERSON
the Centre for Commercialization of Regenerative Medicine ORG
Toronto GPE
two CARDINAL
2019 DATE
Stephen A. Bustin PERSON
PCR ORG
RNA ORG
RNA ORG
10 CARDINAL
PCR ORG
COVID-19 PRODUCT
NB ORG
WHO ORG
the British Medical Journal ORG
2010 DATE
WHO ORG
CDC ORG
RT-PCR ORG
Fatal Therapie WORK OF ART
COVID-19 ORG
```

```
COVID-19 ORG
1] Sensitivity ORG
2 CARDINAL
Thomas Löscher PERSON
March 6, 2020 DATE
3 CARDINAL
Science ORG
July 6, 2001 DATE
24 CARDINAL
4 CARDINAL
Charles Calisher PERSON
May 10, 2020 DATE
5 CARDINAL
Hamburg GPE
Germany GPE
2006 DATE
Virus-Mania ORG
Klaus Kohnlein PERSON
2009 DATE
German NORP
Rubikon PERSON
Süddeutsche Zeitung ORG
Financial Times Deutschland ORG
Konstantin Demeter PERSON
Torsten Engelbrecht PERSON
COVID-19 ORG
Rubikon PERSON
Swiss Italian NORP
Exemple d'affichage graphique des entités nommées d'un texte
nlp = spacy.load("en core web sm")
text = dftrain['text'].iloc[3] # Choisir le premier texte comme
exemple
doc = nlp(text)
displacy.render(doc, style="ent", jupyter=True)
<IPython.core.display.HTML object>
Cette fonction a pour but d'ajouter le type d'entité de chaque token détecté à coté de
chaque entité détéctée dans le texte qu'elle prend en entrée, et retourne le texte modifié
avec les informations ajoutées sur les entités nommées.
def add_entity_name(text):
    if text is None or pd.isna(text):
        return ""
    doc = nlp(text)
    # Créer une liste pour stocker les nouveaux tokens
    nouveaux tokens = []
```

```
# Parcourir les tokens et ajouter des informations selon le type
d'entité
    for token in doc:
        if token.ent type :
            # Ajouter le nom de l'entité (type d'entité) à côté de
l'entité
            nouveaux tokens.append(f"{token.text}
({token.ent type })")
        else:
            nouveaux tokens.append(token.text)
    # Reconstruire le texte avec les informations ajoutées
    nouveau texte = " ".join(nouveaux tokens)
    return nouveau texte
Application de la fonction précédente à la colonne "texte" avec la méthode apply
dftrain['text'] = dftrain['text'].astype(str)
# Appliquer la fonction à la colonne "title"
dftrain['text'] = dftrain['text'].apply(add entity name)
print(dftrain['text'])
615
        It 's been a long time coming , but finally we...
1303
        Constitutional Attorney Matthew (PERSON) DePer...
        The (GPE) United (GPE) States (GPE) is witness...
1232
2022
        After three (DATE) decades (DATE) on the bench...
287
        Based on actual results and accounting for sta...
        5 (MONEY) Million (MONEY) Muslim (MONEY) Child...
1006
1543
        The bombshell claim comes from over (TIME) 20 ...
        BILL (ORG) GATES (ORG) EXPLAINS (ORG) THAT (OR...
853
296
        Let our journalists help you make sense of the...
1325
        Though the whole world relies on RT (ORG) - (0...
Name: text, Length: 844, dtype: object
On scinde les données de la colonne "texte" en jeu d'apprentissage et jeu de test
X=dftrain['text']
X train, X test, y train, y test=train test split(X, y, test size =
0.2, random state=8)
print("X train is \n", X train.shape)
print("y_train is \n",y_train.shape)
print("X test is \n", X_test.shape)
print("y_test is \n",y_test.shape)
X train is
(675,)
y train is
```

```
(675,)
X_test is
(169,)
y_test is
(169,)
```

Ici, c'est une étape importante, on va tester différents classifieurs, pour chacun des classifieurs, on va appliquer le prétraitement + Vectorisation TfIdf, et on applique une cross\_val\_score avec un Kfold de 10 fois, par la suite on stocke dans une liste all\_results la moyenne des accuracy + l'écart type et on la trie par ordre décroissant de moyenne d'accuracy et d'écart type. on remarque que les 2 meilleurs sont SVM et RF qu'on va séléctionner pour leur appliquer le GridSearch sur les paramètres des prétraitements + leurs hyperparamètres pour pouvoir choisir le meilleur.

```
np.random.seed(42) # Set the random seed for NumPy
score = 'accuracy'
seed = 7
allresults = []
results = []
names = []
# Liste des modèles à tester
models = [
    ('MultinomialNB', MultinomialNB()),
    ('LogisticRegression', LogisticRegression(random state=42))
1
models.append(('KNN', KNeighborsClassifier()))
models.append(('CART', DecisionTreeClassifier(random state=42)))
models.append(('RF', RandomForestClassifier(random state=42)))
models.append(('SVM', SVC(random state=42)))
# Création d'un pipeline pour chaque modèle
pipelines = []
for name, model in models:
    pipeline = Pipeline([
        ('normalize', TextNormalizer()),
        ('tfidf', TfidfVectorizer()),
        (name, model)
    1)
    pipelines.append((name, pipeline))
    #pipeline.fit(X train, y train)
all results=[]
scores=[]
for p in pipelines:
    print(p[1])
```

```
# cross validation en 10 fois
    kfold = KFold(n splits=10, random state=seed, shuffle=True)
    print ("Evaluation de ",p)
    start time = time.time()
    # application de la classification
    cv results = cross val score(p[1],X train,y train, cv=kfold,
scoring=score)
    #print("Pour le classifieur",p[0],"on a un score
de",cv results.mean(),"et un écart type de",cv results.std())
    scores.append(cv results)
    all results.append((p[0],cv results.mean(),cv results.std()))
    end time = time.time()
print("all resultats", all results)
all results = sorted(all results, key=lambda x: (-x[1], -x[2]))
print("all resultats", all_results)
    # affichage des résultats
#print ('\nLe meilleur resultat : ',max(results))
X train (675,)
y train (675.)
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('MultinomialNB', MultinomialNB())])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('LogisticRegression',
LogisticRegression(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('KNN', KNeighborsClassifier())])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('CART', DecisionTreeClassifier(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('RF', RandomForestClassifier(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('SVM', SVC(random state=42))])
all resultats [('MultinomialNB', 0.8013388937664618,
0.06343279404927976), ('LogisticRegression', 0.8238147497805093,
0.03841540798676155), ('KNN', 0.6782484635645304, 0.0667016883923831),
('CART', 0.8178226514486392, 0.033587110207881604), ('RF',
0.9008121158911327, 0.03275600442499714), ('SVM', 0.894951712028095,
0.04008495427247786)1
```

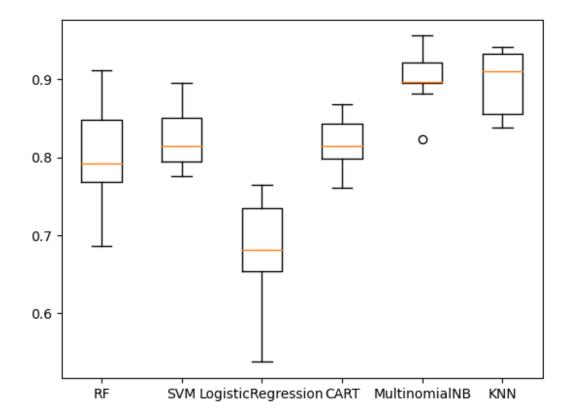
```
all resultats [('RF', 0.9008121158911327, 0.03275600442499714), ('SVM', 0.894951712028095, 0.04008495427247786), ('LogisticRegression', 0.8238147497805093, 0.03841540798676155), ('CART', 0.8178226514486392, 0.033587110207881604), ('MultinomialNB', 0.8013388937664618, 0.06343279404927976), ('KNN', 0.6782484635645304, 0.0667016883923831)]
```

On affiche les accuracy de chaque classifieur, on remarque la médiane (en rouge) de chaque et l'écart type aussi.

```
fig = plt.figure()
fig.suptitle('Comparaison des algorithmes')
ax = fig.add_subplot(111)
plt.boxplot(scores)
ax.set_xticklabels(name[0] for name in all_results)

[Text(1, 0, 'RF'),
    Text(2, 0, 'SVM'),
    Text(3, 0, 'LogisticRegression'),
    Text(4, 0, 'CART'),
    Text(5, 0, 'MultinomialNB'),
    Text(6, 0, 'KNN')]
```

## Comparaison des algorithmes



On a un pipeline pour chaque prétraitement différent, on essaye pas mal (miniscule, lemmatisation, miniscule + lemmatisation..) et on stocke le fit\_transorm de nos X\_train, X\_test sur les pipelines dans des listes qui vont contenir tous les fit\_transform des pipelines pour chaque classifieur, par la suite on parcourt ces listes là, on itère dessus, et chaque élement de la liste (train) va passer par le GridSearch et puis on predict sur son corresapondant dans liste (test).

```
np.random.seed(42) # Set the random seed for NumPy
print("y_train", y_train.shape)
print("y_test", y_test.shape)
X test = np.ravel(X test)
print("X test", X test.shape)
# le plus simple est de faire un test sur differents pipelines.
# pipeline de l'utilisation de CountVectorizer sur le texte avec
differents pre-traitements
CV_brut = Pipeline([('cleaner', TextNormalizer()),
                     ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowcase = Pipeline([('cleaner',
TextNormalizer(removestopwords=False,lowercase=True,
getstemmer=False,removedigit=False)),
                     ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowStop = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=False, removedigit=False)),
                     ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowStopstem = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=True, removedigit=False)),
                     ('count_vectorizer',
CountVectorizer(lowercase=False))])
# pipeline de l'utilisation de TfidfVectorizer avec differents pre-
traitements
TFIDF_brut = Pipeline ([('cleaner', TextNormalizer()),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowcase = Pipeline([('cleaner',
TextNormalizer(removestopwords=False,lowercase=True,
getstemmer=False, removedigit=False)),
```

```
('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowStop = Pipeline([('cleaner'
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=False, removedigit=False)),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowStopstem = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=True, removedigit=False)),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
# Liste de tous les modeles à tester
all models = [
    ("CV_brut", CV_brut),
    ("CV_lowcase", CV_lowcase),
    ("CV_lowStop", CV_lowStop),
    ("CV_lowStopstem", CV_lowStopstem),
    ("TFIDF_lowcase", TFIDF_lowcase),
("TFIDF_lowStop", TFIDF_lowStop),
    ("TFIDF_lowStopstem",TFIDF_lowStopstem),
    ("TFIDF_brut", TFIDF_brut)
1
X train text SVC = []
X_{\text{test\_text\_SVC}} = []
X train text RandomForestClassifier = []
X test text RandomForestClassifier = []
for name, pipeline in all models :
    X train text SVC.append(pipeline.fit transform(X train).toarray())
    X test text SVC.append(pipeline.transform(X_test).toarray())
X train text RandomForestClassifier.append(pipeline.fit transform(X tr
ain).toarray())
X test text RandomForestClassifier.append(pipeline.transform(X test).t
oarray())
```

```
models = {
    'SVC': SVC(random state=42),
    'RandomForestClassifier': RandomForestClassifier(random state=42)
}
params = \{'SVC': [\{'C': [0.001, 0.01, 0.1, 1,2,5,7,10]\},
             {'gamma': [0.001, 0.01, 0.1,0.2,0.3,0.5,0.7,1]},
             {'kernel': ['linear', 'rbf']}],
    'RandomForestClassifier': [{'n estimators': [10, 50, 100, 200,
300]},
                              {'max features': ['auto', 'sqrt',
'log2']}]
for model name, model in models.items():
    score='accuracy'
    X train text = eval('X train text ' + model name)
    X_test_text = eval('X_test_text_' + model_name)
    for i in range (len(X train text)):
      grid search = GridSearchCV(model, params[model name], n jobs=-1,
verbose=1,scoring=score)
      print("grid search fait")
      print("X_train",X_train_text[i].shape)
      print("y_train",y_train.shape)
      grid_search.fit(X_train_text[i],y_train)
      print ('meilleur score %0.3f'%(grid search.best score ),'\n')
      print ('meilleur estimateur',grid_search.best_estimator_,'\n')
      y pred = grid search.predict(X test text[i])
      MyshowAllScores(y test,y pred)
      print("Ensemble des meilleurs paramètres :")
      best parameters = grid search.best estimator .get params()
      for param dict in params[model name]:
        for param name, param value in param dict.items():
            print("\t%s: %r" % (param name,
best parameters[param name]))
y train (675,)
y test (169,)
X test (169,)
grid search fait
X train (675, 24232)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.853
meilleur estimateur SVC(kernel='linear', random state=42)
Accuracy: 0.858
```

```
Classification Report
              precision
                           recall f1-score
                                               support
                0.87342
                          0.83133
                                    0.85185
                                                    83
       FALSE
        TRUE
                0.84444
                          0.88372
                                    0.86364
                                                    86
                                    0.85799
                                                   169
    accuracy
                0.85893
                          0.85752
                                    0.85774
                                                   169
   macro avg
                                    0.85785
weighted avg
                0.85867
                          0.85799
                                                   169
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 'scale'
     kernel: 'linear'
grid search fait
X train (675, 20676)
v train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.853
meilleur estimateur SVC(kernel='linear', random state=42)
Accuracy: 0.888
Classification Report
              precision
                           recall f1-score
                                               support
                0.92105
                          0.84337
                                    0.88050
                                                    83
       FALSE
        TRUE
                0.86022
                          0.93023
                                    0.89385
                                                    86
                                    0.88757
                                                   169
    accuracy
                0.89063
                          0.88680
                                    0.88718
                                                   169
   macro avg
weighted avg
                0.89009
                          0.88757
                                    0.88730
                                                   169
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 'scale'
     kernel: 'linear'
grid search fait
X train (675, 20548)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.846
meilleur estimateur SVC(gamma=0.01, random state=42)
Accuracy: 0.905
Classification Report
              precision recall f1-score
                                               support
```

```
FALSE
                0.83838
                          1.00000
                                    0.91209
                                                    83
        TRUE
                1.00000
                          0.81395
                                    0.89744
                                                    86
                                    0.90533
                                                   169
    accuracy
   macro avg
                0.91919
                          0.90698
                                    0.90476
                                                   169
                0.92063
                          0.90533
                                    0.90463
                                                   169
weighted avg
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 0.01
     kernel: 'rbf'
grid search fait
X train (675, 13959)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.850
meilleur estimateur SVC(kernel='linear', random state=42)
Accuracy: 0.899
Classification Report
              precision
                           recall f1-score
                                               support
       FALSE
                0.94595
                          0.84337
                                    0.89172
                                                    83
        TRUE
                0.86316
                          0.95349
                                    0.90608
                                                    86
                                    0.89941
                                                   169
    accuracy
   macro avg
                0.90455
                          0.89843
                                    0.89890
                                                   169
weighted avg
                0.90382
                          0.89941
                                    0.89903
                                                   169
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 'scale'
     kernel: 'linear'
grid search fait
X train (675, 20676)
y_train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.889
meilleur estimateur SVC(C=1, random state=42)
Accuracy: 0.923
Classification Report
              precision
                           recall f1-score
                                               support
       FALSE
                0.92683
                          0.91566
                                    0.92121
                                                    83
        TRUE
                0.91954
                          0.93023
                                    0.92486
                                                    86
```

```
0.92308
                                                  169
    accuracy
                                    0.92303
                0.92318
                          0.92295
                                                  169
   macro avg
                                    0.92307
weighted avg
                0.92312
                          0.92308
                                                  169
Ensemble des meilleurs paramètres :
     C: 1
     gamma: 'scale'
     kernel: 'rbf'
grid search fait
X train (675, 20548)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.895
meilleur estimateur SVC(C=2, random state=42)
Accuracy: 0.917
Classification Report
              precision
                         recall f1-score
                                              support
       FALSE
                0.88764
                          0.95181
                                    0.91860
                                                   83
        TRUE
                0.95000
                          0.88372
                                    0.91566
                                                   86
                                    0.91716
                                                  169
    accuracy
                          0.91776
   macro avg
                0.91882
                                    0.91713
                                                  169
weighted avg
                0.91937
                          0.91716
                                    0.91711
                                                  169
Ensemble des meilleurs paramètres :
     C: 2
     gamma: 'scale'
     kernel: 'rbf'
grid search fait
X train (675, 13959)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.890
meilleur estimateur SVC(C=2, random state=42)
Accuracy: 0.941
Classification Report
              precision
                           recall f1-score
                                              support
                0.92941
                          0.95181
                                    0.94048
       FALSE
                                                   83
        TRUE
                0.95238
                          0.93023
                                    0.94118
                                                   86
                                    0.94083
                                                  169
    accuracy
                0.94090
                          0.94102
                                    0.94083
                                                  169
   macro avg
```

0.94083

0.94083

169

weighted avg

0.94110

Ensemble des meilleurs paramètres : C: 2 gamma: 'scale' kernel: 'rbf' grid search fait X train (675, 24232) y train (675,) Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.890 meilleur estimateur SVC(C=2, random\_state=42) Accuracy: 0.923 Classification Report precision recall f1-score support FALSE 0.90698 0.93976 0.92308 83 TRUE 0.93976 0.90698 86 0.92308 accuracy 0.92308 169 0.92337 0.92337 0.92308 169 macro avq weighted avg 0.92366 0.92308 0.92308 169 Ensemble des meilleurs paramètres : C: 2 gamma: 'scale' kernel: 'rbf' grid search fait X train (675, 24232) y\_train (675,) Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.864 meilleur estimateur RandomForestClassifier(max\_features='log2', random state=42) Accuracy: 0.864 Classification Report precision recall f1-score support FALSE 0.84091 0.89157 0.86550 83 TRUE 0.88889 0.83721 0.86228 86 0.86391 169 accuracy 0.86439 macro avq 0.86490 0.86389 169

Ensemble des meilleurs paramètres :

0.86532

0.86391

0.86386

169

weighted avg

n\_estimators: 100
max features: 'log2'

grid search fait

X train (675, 20676)

y train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.874

meilleur estimateur RandomForestClassifier(n\_estimators=50,
random state=42)

Accuracy: 0.882

Classification Report

support	f1-score	recall	precision	
83 86	0.87654 0.88636	0.85542 0.90698	0.89873 0.86667	FALSE TRUE
169 169 169	0.88166 0.88145 0.88154	0.88120 0.88166	0.88270 0.88242	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

n\_estimators: 50

max\_features: 'sqrt'

grid search fait

X\_train (675, 20548)

y train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.867

meilleur estimateur RandomForestClassifier(max\_features='log2',
random\_state=42)

Accuracy: 0.893

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.88235 0.90476	0.90361 0.88372	0.89286 0.89412	83 86
accuracy macro avg weighted avg	0.89356 0.89376	0.89367 0.89349	0.89349 0.89349 0.89350	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 100
max\_features: 'log2'

grid search fait

X train (675, 13959)

y train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.864

meilleur estimateur RandomForestClassifier(random state=42)

Accuracy: 0.876

Classification Report

support	f1-score	recall	precision	
83 86	0.86957 0.88136	0.84337 0.90698	0.89744 0.85714	FALSE TRUE
169 169 169	0.87574 0.87546 0.87557	0.87518 0.87574	0.87729 0.87693	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

n\_estimators: 100

max\_features: 'sqrt'

grid search fait

X\_train (675, 20676)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.874

meilleur estimateur RandomForestClassifier(n\_estimators=300, random\_state=42)

Accuracy: 0.929

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.90805 0.95122	0.95181 0.90698	0.92941 0.92857	83 86
accuracy macro avg weighted avg	0.92963 0.93002	0.92939 0.92899	0.92899 0.92899 0.92898	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 300

max features: 'sqrt'

grid search fait

X train (675, 20548)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.877

meilleur estimateur RandomForestClassifier(n\_estimators=200, random state=42)

Accuracy: 0.917

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.90588 0.92857	0.92771 0.90698	0.91667 0.91765	83 86
accuracy macro avg weighted avg	0.91723 0.91743	0.91734 0.91716	0.91716 0.91716 0.91717	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 200
max features: 'sqrt'

grid search fait

X\_train (675, 13959)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.874

meilleur estimateur RandomForestClassifier(n\_estimators=200, random\_state=42)

Accuracy: 0.911

Classification Report

support	f1-score	recall	precision	
83 86	0.91018 0.91228	0.91566 0.90698	0.90476 0.91765	FALSE TRUE
169 169 169	0.91124 0.91123 0.91125	0.91132 0.91124	0.91120 0.91132	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

n\_estimators: 200
max\_features: 'sqrt'

grid search fait

X train (675, 24232)

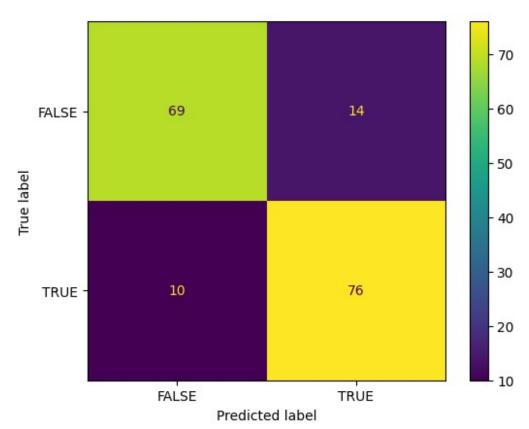
y\_train (675,)

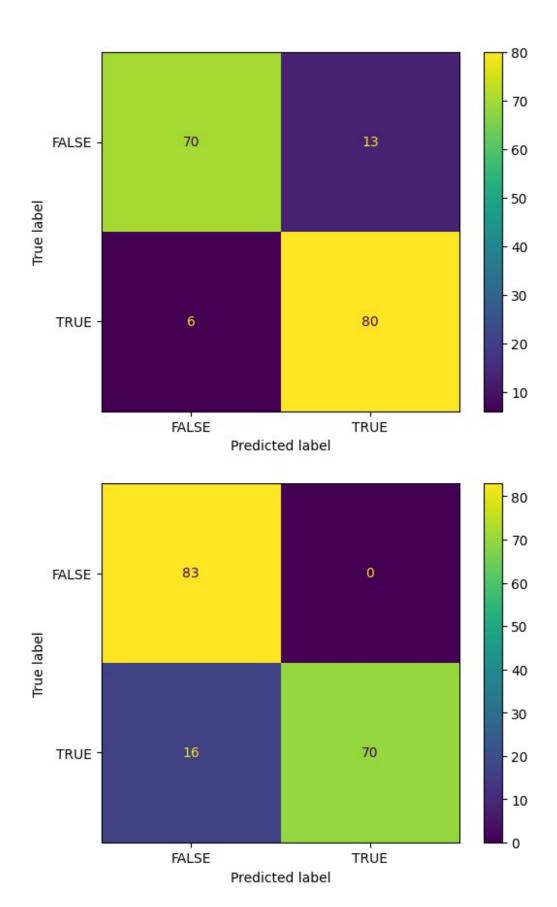
Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.881

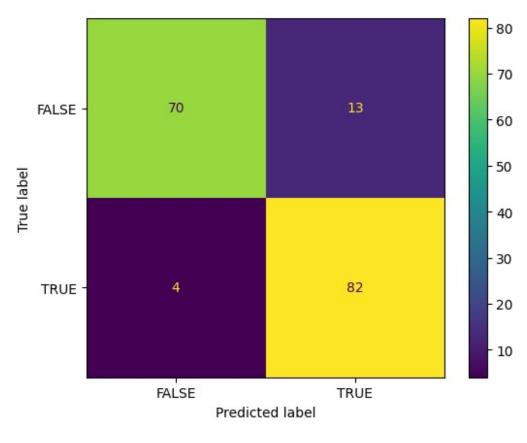
meilleur estimateur RandomForestClassifier(n\_estimators=200, random state=42) Accuracy : 0.905 Classification Report

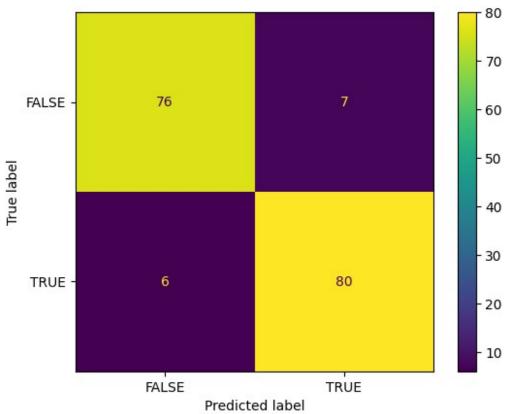
	precision	recall	f1-score	support
FALSE TRUE	0.90361 0.90698	0.90361 0.90698	0.90361 0.90698	83 86
accuracy macro avg weighted avg	0.90530 0.90533	0.90530 0.90533	0.90533 0.90530 0.90533	169 169 169

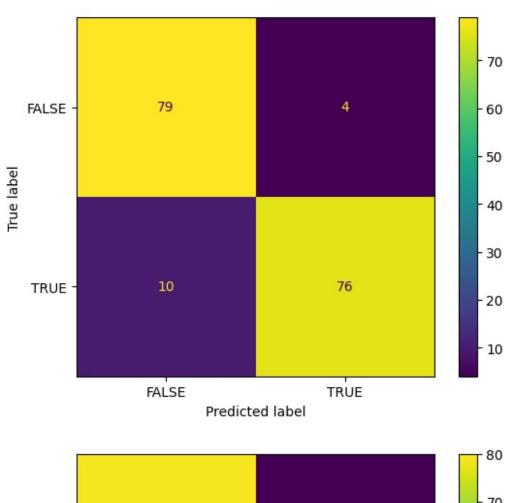
Ensemble des meilleurs paramètres : n\_estimators: 200 max\_features: 'sqrt'

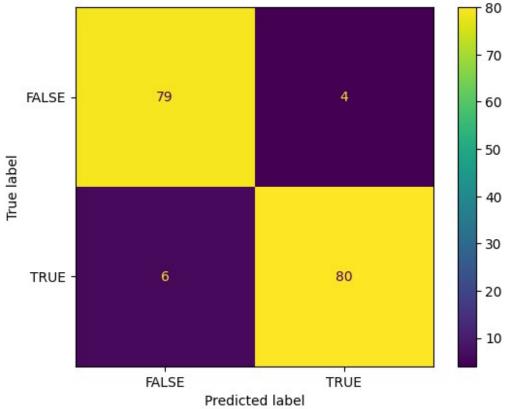


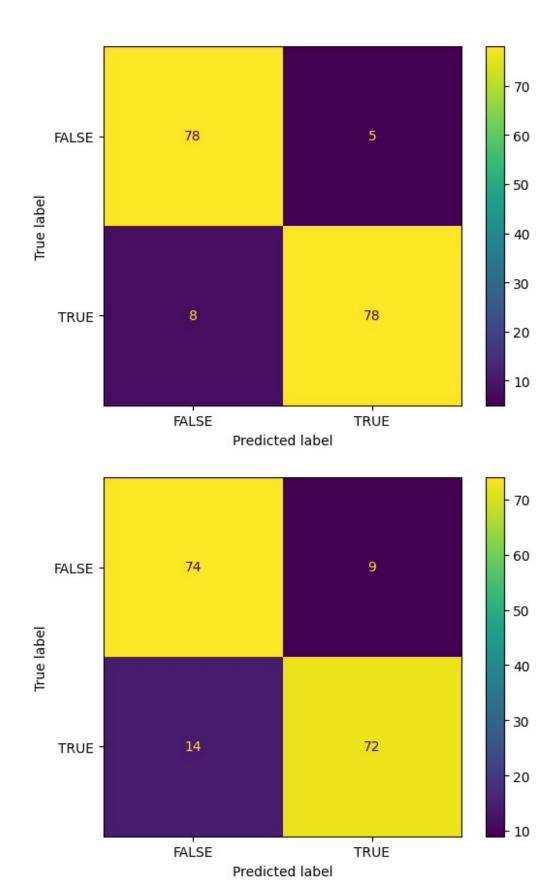


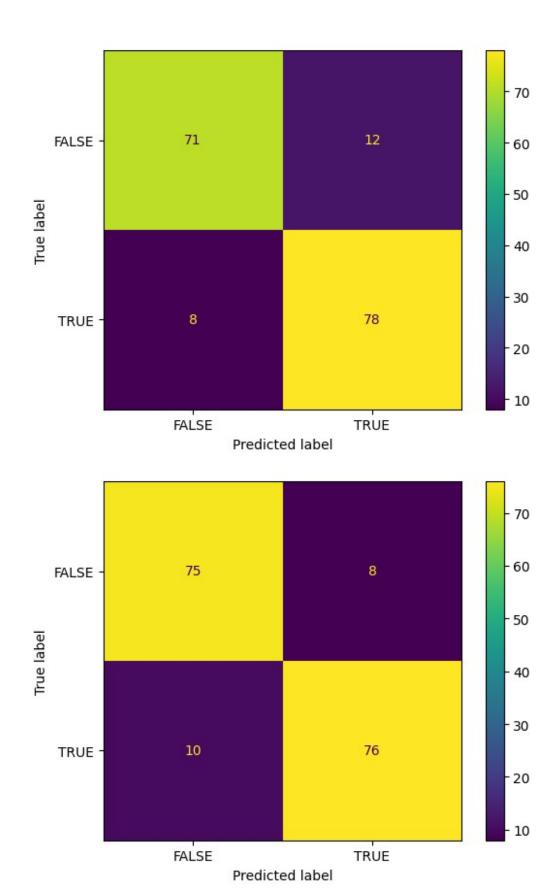


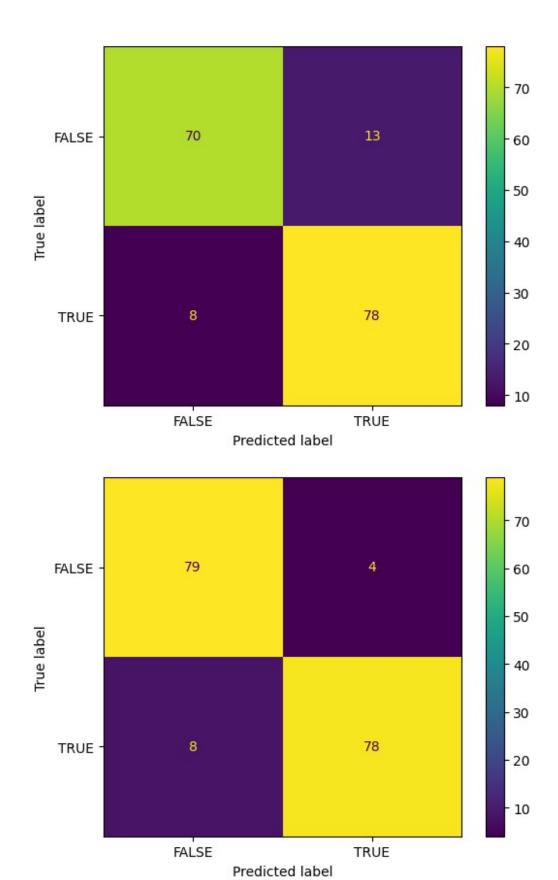


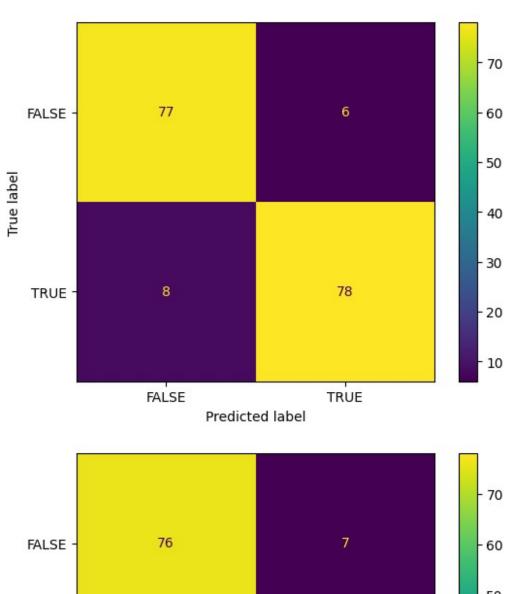


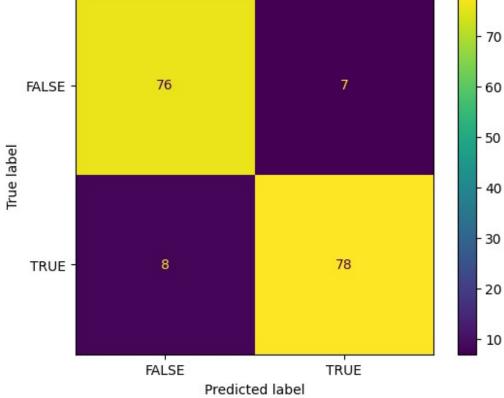


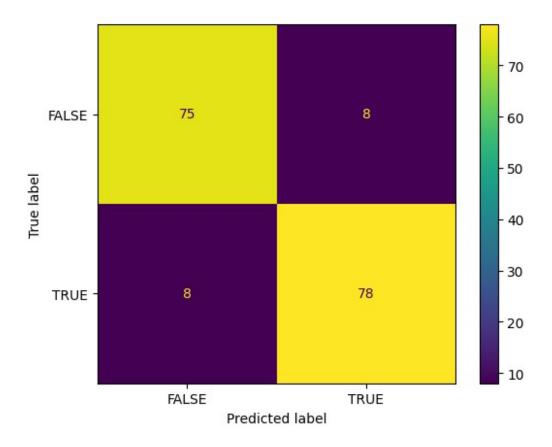












##Etape 3 : Classification selon la colonne TEXT + extraction/application d'entités nommées

Avant de classifier on applique les entités nommées, Cette fonction utilise Spacy pour extraire les entités nommées à partir d'une liste de titres et les stocke dans une liste pour être affichées avec leurs labels correspondants.

```
texte=dftrain['title']
#Créer une liste pour stocker les entités nommées
entities = []
# Traiter chaque texte individuellement et ajouter les entités à la
liste
for phrase in texte:
    phrase_str = str(phrase) # Convertir l'élément en chaîne de
caractères
    doc = nlp(phrase_str)
    for ent in doc.ents:
        entities.append((ent.text, ent.label_))

# Afficher les entités nommées
for entity in entities:
    print(entity[0], entity[1])
```

JK Rowling Confirms Stance Against Transgender Women ORG

MI Sec PERSON

State Official Caught On Video Telling Volunteers ORG

Antrim County GPE

Sarah Parker PERSON

Supreme Court ORG

Trump Leading PERSON

tens of thousands CARDINAL

Britain GPE

Ministry of Ayurveda, Yoga & Naturopathy ORG

Unani PERSON

Siddha PERSON

Homoeopathy PERSON

Ayurveda ORG

Siddha PERSON

Rand Paul PERSON

111 CARDINAL

NHS ORG

Scott Walker PERSON

Mary Burke PERSON

the last century DATE

1918 DATE

Cecile Viboud PERSON

the National Institutes of Health's Fogarty International Center ORG

the "Great Influenza" ORG

an estimated 675,000 CARDINAL

the United States GPE

covid-19 PERSON

10 CARDINAL

covid-19 PERSON

covid-19 DATE

about 6 percent PERCENT

the United States GPE

less than 1 percent PERCENT

Coronavirus Updates ORG

New York GPE

Andrew M. Cuomo PERSON

Monday DATE

New York City GPE

25 percent PERCENT

8.8 million CARDINAL

more than 12,000 CARDINAL

covid-19 PERSON

another 5,000 CARDINAL

between 0.5 CARDINAL

0.8 percent PERCENT

recent weeks DATE

Cuomo PERSON

Monday DATE

New York GPE

0.5 percent PERCENT

one CARDINAL

200 CARDINAL

0.5 percent PERCENT

Viboud PERSON

covid-19 PERSON

more than two CARDINAL

three CARDINAL

Epidemiologists NORP

between 40 to CARDINAL

the next couple of years DATE

Jeffrey Shaman PERSON

Columbia University ORG

Shaman PERSON

only 1 CARDINAL

12 CARDINAL

the United States GPE

0.6 percent PERCENT

New York City GPE

the United States GPE

1 million CARDINAL

half CARDINAL

Shaman PERSON

20 CARDINAL

Monday DATE

covid-19 PERSON

George Gilder PERSON

Kevin McCullough PERSON

Town Hall FAC

last week DATE

Antibody Testing: Proves We've Been Had WORK\_OF\_ART

New York State GPE

7.4% PERCENT

two CARDINAL

first ORDINAL

covid-19 PERSON

second ORDINAL

Viboud PERSON

China GPE

recent days DATE

Santa Clara County GPE

Calif. GPE

Silicon Valley LOC

early April DATE

50 to 85 CARDINAL

between 0.12 CARDINAL

0.2 percent PERCENT

covid-19 PERSON

0.1 percent PERCENT

only half CARDINAL

Viboud PERSON

Shaman PERSON

one-quarter DATE

Santa Clara GPE

Statisticians NORP

Santa Clara GPE

Stanford University ORG

Shaman PERSON

Shaman PERSON

New York City GPE

53,000 CARDINAL

the United States GPE

Lawrence O'Donnell' PERSON

Monday, September 16th, 2013 DATE

£570 billion MONEY

the NHS Long Term Plan LOC

Harry PERSON

Meghan PERSON

2.4 MONEY

At least 13 CARDINAL

Israelis NORP

Pfizer ORG

a month DATE

the US Food and Drug Administration ORG

Israel GPE

20 percent PERCENT

the end of December DATE

Israelis NORP

At least 13 CARDINAL

Pfizer ORG

Israeli NORP

Ynet ORG

the Health Ministry ORG

at least 28 hours TIME

one CARDINAL

Ynet ORG

second ORDINAL

Israel GPE

Health Ministry ORG

second ORDINAL

first ORDINAL

Israeli NORP

Ynet ORG

Galia Rahav PERSON

the Infectious Diseases Unit ORG

Sheba Medical Center ORG

second ORDINAL

first ORDINAL

second ORDINAL

first ORDINAL

Last month DATE

FDA ORG

Bell ORG

four CARDINAL

three CARDINAL

Pfizer PERSON

four CARDINAL

FDA ORG

Bell ORG

US GPE

Bell ORG

the past several weeks DATE

Pfizer ORG

December DATE

UK GPE

first ORDINAL

Mexico GPE

early January DATE

minutes TIME

Portugese NORP

Yahaya Bello PERSON

Labour ORG

100,000 CARDINAL

3 years DATE

OMINOUS ORG

U.S. GPE

1776 DATE

SIGNAL ORG

Delta Force Raids Biden Compound ORG

Ukraine GPE

One CARDINAL

Tetyana Obukhanych ORG

Ph.D. WORK\_OF\_ART

the Rockefeller University ORG

New York GPE

Harvard ORG

British Columbia GPE

Ouebec GPE

2011 DATE

48% PERCENT

PhD WORK OF ART

one CARDINAL

two CARDINAL

a few years ago DATE

American NORP

first ORDINAL

two CARDINAL

11 DATE

Ukraine GPE

PhD WORK OF ART

Google ORG

Barack Obama PERSON

Historical Data ORG

Global Warming ORG

White House ORG

Chicago GPE

Israel GPE

Trump ORG

Lebanese Govt ORG

Confirm ORG

Veterans NORP

The Squad Wants Bible Deemed To Be 'Hate Speech ORG

2040 DATE

Dan Jarvis PERSON

Winter DATE

Rubio Comments ORG

U.S. GPE

NAMES ORG

ON CHINA COMMUNIST PARTY ORG

Biden PERSON

\$3 Trillion MONEY

Plane GPE

One CARDINAL

four CARDINAL

Northern Ireland GPE

Hurricane Patricia EVENT

The Global Warming Scare ORG

Grand Rapids GPE

Ilhan Omar Tells PERSON

Minnesota LOC

Muslim NORP

One CARDINAL

Tilapia ORG

Tilapia ORG

Wake Forest University School of Medicine ORG

Tilapia PERSON

Tilapia ORG

Tilapia ORG

one CARDINAL

Tilapia PERSON

Tilapia ORG

Tilapia PERSON

Tilapia PERSON

Tilapia GPE

Florida GPE

Greenland GPE

annual DATE

Nigeria GPE

Treasury ORG

Russia GPE

Scientific Data ORG

Hoax ORG

a Fake Pandemic WORK OF ART

Boris Johnson PERSON

EU ORG

Meghan Markle PERSON

America GPE

first ORDINAL

Buffalo Cops Charged ORG

Felony Assault for Knocking Down Elderly Man ORG

Vladimir Putin PERSON

Europe LOC

Nicola Sturgeon's ORG

Historical Data ORG

Global Warming ORG

Laura Trott PERSON

Jason Kenny PERSON

Team GB's ORG

Olympics EVENT

Trump Revokes All ORG

First ORDINAL

650 CARDINAL

UK GPE

Unknowns GPE

Egypt GPE

The New York Times ORG

DISGRACEFUL ORG

EU ORG

British NORP

140MILLION MONEY

four million CARDINAL

a year DATE

England GPE

Wales GPE

Wednesday 25 April 2018 DATE

Conor Lamb PERSON

Gingrich Tragedy - PERSON

Broward County Sheriff Fired For Lying About Parkland ORG

The World Health Organization ORG

months DATE

WHO ORG

WHO ORG

34 CARDINAL

Monday October 5th DATE

Michael Ryan PERSON

WHO ORG

roughly 10% PERCENT

around 35 million CARDINAL

Margaret Harris PERSON

WHO ORG

WHO ORG

Ryan PERSON

roughly 7.8 billion CARDINAL

10% PERCENT

780 million CARDINAL

1,061,539 CARDINAL

0.14% PERCENT

0.14% PERCENT

24 CARDINAL

WHO ORG

3.4% PERCENT

March DATE

Covid PERSON

IFR ORG

0.14% PERCENT

Covid PERSON

Dr Ryan's PERSON

WHO ORG

Covid NORP

White House ORG

Chicago GPE

111 CARDINAL

NHS ORG

Fauci Hold PERSON

CDC ORG

Overnight TIME

Ohio Sets Up COVID-19 FEMA Camps ORG

Global Ocean Circulation ORG

Washington State GPE

four million CARDINAL

a year DATE

TEXAS RURAL HOSPITAL ORG

1995-2015 DATE

Putin's End-of-Year Press Conference PERSON

Ouotes GPE

Lawrence O'Donnell' PERSON

Monday, September 16th, 2013 DATE

WADA GPE

Moscow GPE

IOC ORG

Olympics EVENT

DEMOCRATS NORP

Marine Corps ORG

Caucasian Jack PERSON

today DATE

James I about PERSON

four centuries ago DATE

Duane Carr PERSON

the University of Tennessee College of Medicine ORG

Victor Buhler PERSON

St. Joseph Hospital FAC

Kansas City GPE

thousands CARDINAL

Sheldon Sommers PERSON

Pathologist NORP

Lenox Hill Hospital ORG

New York GPE

Brigitte ORG

Youtube PERSON

German NORP

Rich White's PERSON

Smoke Screens PERSON

Jan Zeldenrust PERSON

Dutch NORP

the Government of Holland ORG

1951 - 1984 DATE

1980 DATE

Dutch NORP

these days DATE

Coal ORG

First ORDINAL

emphysema PERSON

4 CARDINAL

5 CARDINAL

4 CARDINAL

8) CARDINAL

first ORDINAL

emphysema PERSON

secondly ORDINAL

emphysema PERSON

Delhi GPE

'Delhi GPE

Delhi GPE

Indian NORP

Delhi GPE

Indian NORP

Delhi GPE

'Delhi GPE

emphysema PERSON

one CARDINAL

BioFlex ORG

firstly ORDINAL

secondly ORDINAL

the University of Arkansas ORG

Villanova University ORG

Marquette University ORG

more than 500 CARDINAL

U.S. GPE

Canadian NORP

Scot Burton PERSON

the Sam M. Walton College of Business ORG

two CARDINAL

The New York Times ORG

5 CARDINAL

American NORP

5 CARDINAL

Vanuatu ORG

Cyclone Pam PERSON

Wisconsin GPE

Newly ORG

the Department of Instruction ORG

the past year DATE

Wisconsin GPE

2nd DATE

41st ORDINAL

more than half CARDINAL

today DATE

the Walker Administration ORG

Democratic Party of Wisconsin Chair ORG

Martha Laning PERSON

Thursday DATE

last year DATE

State ORG

Scott Walker PERSON

Wisconsin GPE

second ORDINAL

five days DATE

Walker's ORG

State ORG

the last six years DATE

36 CARDINAL

22.0 CARDINAL

20.0 CARDINAL

Wisconsin GPE

Midwest LOC

Kentucky GPE

41st ORDINAL

Wisconsin GPE

Wisconsin GPE

fourth ORDINAL

K-12 PRODUCT

2015-16 DATE

the Center on Budget and Policy Priorities ORG

Wisconsin GPE

Wisconsin GPE

Midwest LOC

ten CARDINAL

Wisconsin GPE

Republican NORP

EU ORG

McFaul PERSON

Russian NORP

Putin PERSON

NBA Players Wear Special Lace Collars To Honor Ruth Bader Ginsburg ORG

Ghislaine Maxwell PERSON

COVID-19 ORG

New Hampshire GPE

NHS ORG

Marco Rubio Just Publicly Shamed His Own Constituents For Flocking

**PERSON** 

ACCLAIMED ISRAELI ASTROPHYSICIST ORG

Jeremy Corbyn Vows PERSON

George Soros PERSON

Nicola Sturgeon's ORG

five CARDINAL

Ireland GPE

one CARDINAL

two CARDINAL

Obama PERSON

November 14, 2014 DATE

Fastest Rate FAC

Last 28 Centuries DATE

Trump ORG

thousands CARDINAL

Wisconsin GPE

the Russian Federation GPE

the United Nations ORG

Justine Greening PERSON

CDC ORG

COVID-19 ORG

U.S. GPE

CDC ORG

between March and December 2020 DATE

1.5 percent PERCENT

two months DATE

0.5 percent PERCENT

the first 20 days DATE

less than 2 percent PERCENT

100 days DATE

CDC ORG

Vanuatu ORG

Cyclone Pam PERSON

Australian NORP

Paid Liberal Trolls of America ORG

6 hours TIME

30 MONEY

40 MONEY

70 cents MONEY

1 MONEY

Soros PERSON

Joe Barron PERSON

George Soros PERSON

500 MONEY

Art Tubolls PERSON

163 CARDINAL

Soros PERSON

Deluth GPE

Charlie Kirk PERSON

Turning Point PERSON

AAA ORG

Oregon GPE

millions CARDINAL

1 CARDINAL

Hillary Caught PERSON

Kamala Harris PERSON

United States GPE

UK GPE

1,783 CARDINAL

690,773 CARDINAL

the year to March DATE

British NORP

13,000 CARDINAL

England GPE

2010 DATE

April DATE

May this year DATE

3,264 CARDINAL

NMC ORG

more than 4,500 CARDINAL

UK GPE

Saffron Cordery ORG

NHS Providers ORG

NHS ORG

NHS ORG

**Brexit PERSON** 

UK GPE

NHS ORG

AAA ORG

Oregon GPE

millions CARDINAL

1 CARDINAL

Conor Lamb PERSON

B.C. GPE

ALS ORG

years DATE

24-hour TIME

More Than 10,000 CARDINAL

Cast Ballots PERSON

Michigan GPE

Miles PERSON

1984 DATE

the New York Times ORG

Trump ORG

NHS ORG

30% PERCENT

today DATE

James I about PERSON

four centuries ago DATE

Duane Carr PERSON

the University of Tennessee College of Medicine ORG

Victor Buhler PERSON

St. Joseph Hospital FAC

Kansas City GPE

thousands CARDINAL

Sheldon Sommers PERSON

Pathologist NORP

Lenox Hill Hospital ORG

New York GPE

Brigitte ORG

Youtube PERSON

German NORP

Rich White's PERSON

Smoke Screens PERSON

Jan Zeldenrust PERSON

Dutch NORP

the Government of Holland ORG

1951 - 1984 DATE

1980 DATE

Dutch NORP

these days DATE

Coal ORG

First ORDINAL

emphysema PERSON

4 CARDINAL

5 CARDINAL

4 CARDINAL

8) CARDINAL

first ORDINAL

emphysema PERSON

secondly ORDINAL

emphysema PERSON

Delhi GPE

'Delhi GPE

Delhi GPE

Indian NORP

Delhi GPE

Indian NORP

Delhi GPE

'Delhi GPE

emphysema PERSON

one CARDINAL

BioFlex ORG

firstly ORDINAL

secondly ORDINAL

the University of Arkansas ORG

Villanova University ORG

Marguette University ORG

more than 500 CARDINAL

U.S. GPE

Canadian NORP

Scot Burton PERSON

the Sam M. Walton College of Business ORG

two CARDINAL

The Prices Of Food Items Are Coming Down ORG

Garba Shehu PERSON

Belgian NORP

Fremault PERSON

Brussels GPE

5 CARDINAL

Celine Fremault PERSON

the Brussels-Capital Region Responsible For Housing, Quality Of Life,

Environment And Energy ORG

5 CARDINAL

Brussels GPE

Fremault PERSON

5 CARDINAL

Belgian NORP

9 CARDINAL

95 mW/ QUANTITY

m2 PERSON

1 mW/m2 QUANTITY

thousands CARDINAL

Brussels GPE

EU ORG

one CARDINAL

2 CARDINAL

5 CARDINAL

EU ORG

EU ORG

Celine Fremault PERSON

North America LOC

Celine Fremault PERSON

Brussel GPE

5 CARDINAL

6 minute TIME

UN ORG

5 CARDINAL

zero CARDINAL

Tom Wheeler PERSON

FCC Chair ORG

```
5 CARDINAL
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5 CARDINAL

5 CARDINAL

5 CARDINAL

US GPE

Hundreds CARDINAL Martin Pall PERSON

5 CARDINAL

Earth LOC

5 CARDINAL

AI ORG

5 CARDINAL

Brussels GPE

first ORDINAL

5 CARDINAL

dozens CARDINAL

5 CARDINAL

21 CARDINAL

USA GPE

March 13, 2019 DATE

Portland GPE

FCC ORG

EMF Safety Network ORG

WhatIs5G.info ORG

5 CARDINAL

TRUMP ORG

Obama GPE

Reagan PERSON

Same Point ORG

Presidencies FAC

British Government Study Confirms Covid-19 Vaccine Risk: Infections

**ORG** 

Fortnight GPE

January DATE

80 CARDINAL

Pemberley House ORG

Basingstoke GPE

England GPE

a third CARDINAL

24 CARDINAL

Germany GPE

British NORP

Gibraltar GPE

Pfizer ORG

16th January DATE

33 thousand CARDINAL

2761 CARDINAL

9 CARDINAL

December DATE

almost a hundred CARDINAL

The United Arab Emirates GPE

Chinese NORP

Sinopharm ORG

Pfizer ORG

December DATE

Amer Sharif PERSON

the Covid-19 Command & Control Centre ORG

Dubai GPE

UAE ORG

first ORDINAL

100% PERCENT

10th February DATE

daily DATE

18 CARDINAL

first ORDINAL

last spring DATE

Israel GPE

19th December DATE

New York Times ORG

Alex Berenson PERSON

Israelis NORP

January DATE

Palestine GPE

December DATE

Palestinians NORP

Israeli NORP

Hervé Seligmann PERSON

Aix-Marseilles University ORG

about 40 CARDINAL

the three weeks DATE

first ORDINAL

second ORDINAL

Israel GPE

Pfizer ORG

the first week DATE

Raine PERSON

UK GPE

MHRA ORG

two CARDINAL

first ORDINAL

The Public Health England ORG

Matt Hancock PERSON

80% PERCENT

48% PERCENT

first ORDINAL

Pfizer ORG

Astra ORG

Zeneca NORP

Mary Ramsay PERSON

PHE ORG

hundreds CARDINAL

British NORP

the Bill & Melinda Gates Foundation LAW

Covid PERSON

Covid PERSON

Commission ORG

Hungary GPE

Viktor Orban PERSON

Referendum GPE

October 2 DATE

two weeks ago DATE

Jean Claude Juncker PERSON

Governments PERSON

EU ORG

solidarity ORG

Tory PERSON

Steve Baker PERSON

Jean Claude Juncker PERSON

democratic NORP

the European Union ORG

The Greatest Hoax Ever Perpetrated on an Unsuspecting Public

WORK OF ART

Ohio GPE

Tesco ORG

119 CARDINAL

last year DATE

England GPE

Wales GPE

This Winter DATE

Meets With PERSON

GOP ORG

1984 DATE

the New York Times ORG

Trump ORG

Barnardo ORG

Coronavirus LOC

44% PERCENT

Marijuana PERSON

Normandy Four PERSON

State ORG

Democrats NORP

Two CARDINAL

Georgia GPE

Dan Jarvis PERSON

Winter DATE

Massive International Study Shows Countries ORG

79% PERCENT

Lower Mortality Rate ORG

100,000 CARDINAL

American NORP

Trump ORG

Supreme Court ORG

Obamacare NORP

50,000 CARDINAL

Chinese NORP

The New York Times Affirms ORG

Capitol FAC

Legislature ORG

over 200,000 CARDINAL

NHS ORG

2010 DATE

£330 million MONEY

Biden PERSON

the Second Amendment LAW

May 2018 DATE

fourth ORDINAL

the month of May DATE

NOAA ORG

1880 DATE

March-May DATE

fourth ORDINAL

monthly DATE

NOAA ORG

National Centers for Environmental Information ORG

NOAA ORG

May 2018 DATE

May DATE

1.44 CARDINAL

F PRODUCT

the 20th century DATE

58.6 CARDINAL

F PRODUCT

fourth ORDINAL

May DATE

1880-2018 DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.58 CARDINAL

May 2018 DATE

the 42nd consecutive May DATE

the 401st consecutive month DATE

the 20th century DATE

North America LOC

Europe LOC

Asia LOC

Atlantic LOC

Pacific LOC

Indian Oceans NORP

May DATE

Canada GPE

Atlantic Ocean LOC

Greenland GPE

May DATE

2.05 CARDINAL

the 20th century DATE

52.0 CARDINAL

F PRODUCT

May DATE

2011 DATE

2013 DATE

seventh ORDINAL

May DATE

139-year DATE

May 2018 DATE

U.S. GPE

Europe LOC

5.4 CARDINAL

Canada GPE

Russia GPE

5.4 CARDINAL

Europe LOC

May DATE

continental ORG

1910 DATE

4.97 CARDINAL

2003 DATE

1.66 CARDINAL

F PRODUCT

May 2018 DATE

first ORDINAL

May DATE

3.6 CARDINAL

F PRODUCT

2.0 CARDINAL

European NORP

May DATE

Asia LOC

May DATE

2009 DATE

May DATE

1.19 CARDINAL

the 20th century DATE

monthly DATE

61.3 CARDINAL

fourth ORDINAL

May DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.39 CARDINAL

May 2018 DATE

May DATE

Arctic LOC

second ORDINAL

40-year DATE

420,000 square miles QUANTITY

8.1 percent PERCENT

1981-2010 DATE

the National Snow and Ice Data Center(link ORG

NOAA ORG

NASA ORG

May Arctic LOC

2016 DATE

Arctic LOC

Bering GPE

Chukchi Seas PERSON

May DATE

330,000 square miles QUANTITY

8.6 percent PERCENT

1981-2010 DATE

third ORDINAL

May DATE

Antarctic sea LOC

May DATE

the Weddell Sea LOC

Seasonal DATE

March-May 2018 DATE

March-May DATE

1.48 CARDINAL

the 20th century DATE

56.7 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

1880-2018 DATE

0.45 CARDINAL

F cooler PRODUCT

2016 DATE

March-May DATE

2014 DATE

three-month DATE

Europe LOC

the Middle East LOC

Africa LOC

China GPE

Mongolia GPE

South America LOC

Atlantic LOC

Indian NORP

Pacific Oceans NORP

March-May 2018 DATE

March-May DATE

2.34 CARDINAL

the 20th century DATE

46.4 CARDINAL

F PRODUCT

fifth ORDINAL

March-May DATE

139-year DATE

Five CARDINAL

March-May 2018 DATE

nine CARDINAL

continental ORG

1910 DATE

Europe LOC

South America LOC

third ORDINAL

March-May DATE

March-May DATE

1.15 CARDINAL

the 20th century DATE

61.0 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

Year-to-Date DATE

January-May 2018 DATE

1.39 CARDINAL

the 20th century DATE

55.5 CARDINAL

F PRODUCT

2010 DATE

fourth ORDINAL

January-May DATE

139-year DATE

2018 year DATE

0.59 CARDINAL

2016 DATE

2.18 CARDINAL

the 20th century DATE

42.8 CARDINAL

F PRODUCT

2002 DATE

fifth ORDINAL

January-May DATE

1.08 CARDINAL

the 20th century DATE

60.8 CARDINAL

fifth ORDINAL

January-May DATE

1880-2018 DATE

Wisconsin GPE

Laura Trott PERSON

Jason Kenny PERSON

Team GB's ORG

Olympics EVENT

this week DATE

Scottish National Party ORG

Amazon ORG

Obama ORG

BRAZIL GPE

AstraZeneca ORG

Oxford University ORG

Anvisa ORG

one CARDINAL

three CARDINAL

Brazil GPE

Oxford University ORG

the UK Government ORG

100 million CARDINAL

US GPE

September 6 DATE

India GPE

South Africa GPE

AstraZeneca ORG

Europe LOC

Inclusive Vaccine Alliance ORG

400 million CARDINAL

European NORP

UK GPE

30 million CARDINAL

BioNtech/Pfizer ORG

60 million CARDINAL

Valneva - in PERSON

six CARDINAL

UK GPE

Patrick Vallance PERSON

next spring DATE

the Commons and Lords National Security Strategy Committee ORG

AstraZeneca ORG

the turn of the year DATE

UK GPE

Vaccine Taskforce FAC

Kate Bingham PERSON

UK GPE

four CARDINAL

six CARDINAL

three CARDINAL

one CARDINAL

two CARDINAL

Covid PERSON

26,688 CARDINAL

UK GPE

789,229 CARDINAL

191 CARDINAL

today DATE

Fastest Rate FAC

Last 28 Centuries DATE

MSM ORG

Italy GPE

Nazi NORP

pro-Russian NORP

British NORP

Brexit Britain PERSON

EU ORG

SURGES ORG

month DATE

Corbyn PERSON

The New York Times Affirms ORG

Trump ORG

Supreme Court ORG

Obamacare NORP

San Diego GPE

Democrat NORP

Peters PERSON

Trump ORG

Mexico GPE

5 CARDINAL

Miami GPE

Unknowns GPE

Egypt GPE

The New York Times ORG

Carnival Cruise Lines ORG

Biden PERSON

Russia GPE

SolarWinds ORG

N.Y. GPE

Rick Perry PERSON

CANCER IS CANDIDA/FUNGUS AND CAN BE CURED ORG

five CARDINAL

Vigil PERSON

13-year-old DATE

Pittsburg Area Antifa LOC

Tweets Warning PERSON

Trump WORK OF ART

Sunday DATE

noon TIME

2055 CARDINAL

5 CARDINAL

5 CARDINAL

NYC ORG

ICU ORG

Cameron Kyle-Sidell PERSON

tens of thousands CARDINAL

New Yorkers NORP

30,000 feet QUANTITY

30% PERCENT

The Global Warming Scare ORG

N.Y. GPE

A decade DATE

Scottish National Party ORG

Capitol Shooting Exposed ORG

Hollywood GPE

Wisconsin GPE

Newly ORG

the Department of Instruction ORG

the past year DATE

Wisconsin GPE

2nd DATE

41st ORDINAL

more than half CARDINAL

today DATE

the Walker Administration ORG

Democratic Party of Wisconsin Chair ORG

Martha Laning PERSON

Thursday DATE

last year DATE

State ORG

Scott Walker PERSON

Wisconsin GPE

second ORDINAL

five days DATE

Walker's ORG

State ORG

the last six years DATE

36 CARDINAL

22.0 CARDINAL

20.0 CARDINAL

Wisconsin GPE

Midwest LOC

Kentucky GPE

41st ORDINAL

Wisconsin GPE

Wisconsin GPE

fourth ORDINAL

K-12 PRODUCT

2015-16 DATE

the Center on Budget and Policy Priorities ORG

Wisconsin GPE

Wisconsin GPE

Midwest LOC

ten CARDINAL

Wisconsin GPE

Republican NORP

Bill Gates PERSON

COVID-19 Vaccine PERSON

Nearly 1,000,000 CARDINAL

Britons GPE

**Brexit PERSON** 

Illinois General Assembly ORG

NHS ORG

FAKED DATE

Nancy PERSON

These United States GPE

El Niño ORG

2015 DATE

Apocalyptic Sea-Level Rise ORG

Tlaib Files Lawsuit ORG

Alabama Supreme Court Recognizes Unborn ORG

Pakistani NORP

Meesha Shafi PERSON

3 years DATE

Ali Zafar PERSON

Thousands CARDINAL

New Jersey Sept GPE

14, 2020 DATE

Videos GPE

UFO Sighting News ORG

Feds Handcuff 'Irrational' House PERSON

Raiding Her Office ORG

Rastafarian Prez PERSON

Nana Akufo-Addo ORG

FAKED DATE

Pope PERSON

80 CARDINAL

Child Trafficking, Fraud — Conservative WORK OF ART

NHS ORG

1bn ORDINAL

June DATE

Earth LOC

European NORP

Tony Abbott PERSON

the Russian Federation GPE

the United Nations ORG

Central Europe LOC

**Brexit PERSON** 

Britain GPE

Slovak GPE

TEXAS RURAL HOSPITAL ORG

1995-2015 DATE

Pennsylvania GPE

Files Memorandum PERSON

2020 DATE

Tom Hanks PERSON

65 DATE

Passes Away PERSON

Australia GPE

Great Barrier Reef WORK OF ART

95-Degree Days DATE

Google ORG

Adam Hasner PERSON

COVID ORG

11-year-olds ORDINAL

Detroit GPE

Voter Fraud PERSON

At least 13 CARDINAL

Israelis NORP

Pfizer ORG

a month DATE

the US Food and Drug Administration ORG

Israel GPE

20 percent PERCENT

the end of December DATE

Israelis NORP

At least 13 CARDINAL

Pfizer ORG

Israeli NORP

Ynet ORG

the Health Ministry ORG

at least 28 hours TIME

one CARDINAL

Ynet ORG

second ORDINAL

Israel GPE

Health Ministry ORG

second ORDINAL

first ORDINAL

Israeli NORP

Ynet ORG

Galia Rahav PERSON

the Infectious Diseases Unit ORG

Sheba Medical Center ORG

second ORDINAL

first ORDINAL

second ORDINAL

first ORDINAL

Last month DATE

FDA ORG

Bell ORG

four CARDINAL

three CARDINAL

Pfizer PERSON

four CARDINAL

FDA ORG

Bell ORG

US GPE

Bell ORG

the past several weeks DATE

Pfizer ORG

December DATE

UK GPE

first ORDINAL

Mexico GPE

early January DATE

minutes TIME

Portugese NORP

OMINOUS ORG

U.S. GPE

Brexit Britain PERSON

EU ORG

SURGES ORG

FDA Shocking Study: Cells Used ORG

Rick Scott PERSON

Democrats NORP

Liberal States GPE

Kamala Harris Details on How Their Ancestors Owned Slaves ORG

The Unvaccinated Amish Rarely Get Cancer ORG

Britons GPE

**Brexit PERSON** 

Antarctica LOC

Antarctica LOC

Ireland GPE

one CARDINAL

two CARDINAL

Jon Evans PERSON

Up to a third CARDINAL

25 years DATE

Superbowl Champions Refuse Invitation ORG

White House ORG

Commission ORG

Hungary GPE

Viktor Orban PERSON

Referendum GPE

October 2 DATE

two weeks ago DATE

Jean Claude Juncker PERSON

Governments PERSON

EU ORG

solidarity ORG

Tory PERSON

Steve Baker PERSON

Jean Claude Juncker PERSON

democratic NORP

the European Union ORG

AUTOPSY ORG

Chadwick Boseman PERSON

Toronto GPE

Today DATE

USCIS ORG

COVID ORG

Pentagon ORG

Bail Hearing PERSON

Ocala GPE

Florida GPE

Romney Endorses PERSON

Joe Biden PERSON

Alex Trebek Becomes PERSON

U.S. GPE

Google ORG

Israeli NORP

first ORDINAL

year DATE

Dan Aridor PERSON

Accelerated Evolution Biotechnologies Ltd. ORG

2000 DATE

ITEK GPE

SoAP ORG

day one DATE

a few weeks DATE

Aridor PERSON

Ilan Moran PERSON

Ilan Moran PERSON

an estimated 18.1 million MONEY

each year DATE

the International Agency for Research on Cancer ORG

sixth ORDINAL

second ORDINAL

second ORDINAL

Aridor ORG

Ilan Morad PERSON

SoAP ORG

2018 DATE

the Nobel Prize WORK\_OF\_ART

two CARDINAL

Morad GPE

first ORDINAL

Morad GPE

Morad GPE

Hanan Itzhaki PERSON

Morad GPE

at least three CARDINAL

Morad GPE

Morad GPE

three CARDINAL

three CARDINAL

Morad GPE

Morad PERSON

Morad PERSON

12 CARDINAL

Morad PERSON

one CARDINAL

two CARDINAL

Morad GPE

Today DATE

two CARDINAL

Morad GPE

only a few weeks DATE

Aridor GPE

Morad PERSON

first ORDINAL

a few years DATE

four million CARDINAL

a year DATE

Kremlin ORG

Putin PERSON

Russia GPE

Trump ORG

McCain ORG

the South Pole Is Warming LOC

Quickly PERSON

Lee Majors Dead PERSON

83;\$6 Million CARDINAL

2020 DATE

CLARKSBURG ORG

Jared Henry ORG

Maddie Mansell PERSON

Buckhannon Pike PERSON

Nutter Fort GPE

Mansell PERSON

Henry PERSON

Mansell PERSON

Mansell PERSON

Nikki Bowmar PERSON

the U.S. Small Business Administration ORG

Bowmar PERSON

approximately 114,391 CARDINAL

West Virginia GPE

99 percent PERCENT

280,213 CARDINAL

nearly 50 percent PERCENT

Bowmar PERSON

Mansell PERSON

first ORDINAL

Christ GPE

George "Gib" Brown PERSON

The James and Law Co. ORG

West Virginia GPE

nearly 30 CARDINAL

Clarksburg GPE

two CARDINAL

1903 DATE

Brown PERSON

James and Law ORG

Brown PERSON

the years DATE

Chamber ORG

the Robinson Grand Performing Arts Center ORG

Clarksburg GPE

Clarksburg GPE

years DATE

Brown PERSON

Dan Taylor PERSON

the West Virginia Community Development Hub's Energizing

Entrepreneurial Communities ORG

the last of three years DATE

four CARDINAL

Grafton ORG

Taylor PERSON

Taylor PERSON

Stephen Mann PERSON

Lewis County GPE

December DATE

Weston GPE

Mann ORG

two CARDINAL

one CARDINAL

401k PRODUCT

Mann PERSON

Manns PERSON

Lewis County GPE

Mann PERSON

Lewis County GPE

West Virginia GPE

Bowmar PERSON

Mann ORG

The Energizing Entrepreneurial Communities ORG

Grafton ORG

Taylor PERSON

All Aboard Grafton FAC

Taylor PERSON

Taylor County PERSON

first ORDINAL

February DATE

Patricia Henderson PERSON

the Taylor County Development Authority ORG

the Taylor County Commission ORG

Taylor County GPE

Weston GPE

Fairs and Festivals Committee ORG

Julia Spelsberg PERSON

Spelsberg PERSON

Weston GPE

North Central West Virginia LOC

The Small Business Administration ORG

**Bowmar PERSON** 

USDA ORG

the Tax Cut FAC

Taylor PERSON

Northern Ireland GPE

Chinese NORP

CDC ORG

Kremlin ORG

Putin PERSON

Russia GPE

163 CARDINAL

George Soros For PERSON

Clinton Foundation ORG

2055 CARDINAL

Florida GPE

11-year-olds ORDINAL

Italy Arrests Cardinals And Seizes ORG

650 CARDINAL

UK GPE

Bloomberg PERSON

CDC ORG

COVID-19 ORG

U.S. GPE

CDC ORG

between March and December 2020 DATE

1.5 percent PERCENT

two months DATE

0.5 percent PERCENT

the first 20 days DATE

less than 2 percent PERCENT

100 days DATE

CDC ORG

Malaria GPE

The Squad Wants Bible Deemed To Be 'Hate Speech ORG

Trudeau Bought PERSON

the War on Drugs EVENT

Biden PERSON

Russia GPE

SolarWinds ORG

HPV ORG

NY Post's 'Smoking Gun ORG

Hunter Biden PERSON

Forensic Analysis Concludes ORG

David Davis PERSON

Canada GPE

Tusk PERSON

Gohmert Responds PERSON

Trump ORG

McCain ORG

United Airlines ORG

9 CARDINAL

Buhari To Resign PERSON

May 2023 DATE

Lai Mohammed PERSON

2015 DATE

Europe LOC

Spain GPE

France GPE

Germany GPE

Belgium GPE

the Netherlands GPE

Sweden GPE

Sunday DATE

Sweden GPE

100,000 CARDINAL

56.40 CARDINAL

U.K. GPE

69.60 CARDINAL

Spain GPE

60.88 CARDINAL

Italy GPE

58.16 CARDINAL

Sunday DATE

Johns Hopkins University ORG

U.K. GPE

fourth ORDINAL

Spain GPE

Italy GPE

sixth ORDINAL

eighth ORDINAL

Europe LOC

two CARDINAL

Democrat Fundraising Machine 'ActBlue Texas ORG

Caught Paying Out Small Sums WORK OF ART

Hundreds of Individuals for Unknown Purposes - David Harris Jr ORG

Donald Trump PERSON

anti-American NORP

Kenosha GPE

Blake or Rittenhouse ORG

Global Warming PERSON

75 CARDINAL

MPH ORG

Central Europe LOC

Brexit PERSON

Britain GPE

Slovak GPE

Trump Leading PERSON

Italy Arrests Cardinals And Seizes ORG

Travis Rowley PERSON

Democrats NORP

Kamala Harris Details on How Their Ancestors Owned Slaves ORG

one hour TIME

Biden PERSON

the World Health Organization ORG

PCR ORG

WHO ORG

today DATE

COVID ORG

WHO ORG

COVID ORG

March 2020 DATE

WHO ORG

COVID ORG

COVID ORG

Nigeria GPE

hundreds CARDINAL

Catholics NORP

Muslims NORP

Paul Bettencourt PERSON

Wednesday 25 April 2018 DATE

NHS ORG

1bn ORDINAL

Harry PERSON

Meghan PERSON

2.4 MONEY

Grand Rapids GPE

Britain GPE

Georgia GPE

GOP Areas ORG

**DEM Areas PERSON** 

Tony Abbott PERSON

NBA Players Wear Special Lace Collars To Honor Ruth Bader Ginsburg ORG Emmett Till's Memorial Sign Shot PERSON

35 days DATE

Bill Withers PERSON

\$2 Million MONEY

Trump Re-election Campaign ORG

New Yorkers NORP

Government ORG

This Winter DATE

Fauci Hold PERSON

US GPE

CDC ORG

Josh Mandel PERSON

British NORP

100 years ago DATE

Wear a Mask WORK OF ART

78 Year Old DATE

CIA ORG

Disneyland FAC

California GPE

Texas GPE

Another Violent Leftist Arrested ORG

US Capital Protest ORG

Democrat NORP

34-Year-Old DATE

New York GPE

El Nino ORG

Wisconsin GPE

today DATE

James I about PERSON

four centuries ago DATE

Duane Carr PERSON

the University of Tennessee College of Medicine ORG

Victor Buhler PERSON

St. Joseph Hospital FAC

Kansas City GPE

thousands CARDINAL

Sheldon Sommers PERSON

Pathologist NORP

Lenox Hill Hospital ORG

New York GPE

Brigitte ORG

Youtube PERSON

German NORP

Rich White's PERSON

Smoke Screens PERSON

Jan Zeldenrust PERSON

Dutch NORP

the Government of Holland ORG

1951 - 1984 DATE

1980 DATE

Dutch NORP

these days DATE

Coal ORG

First ORDINAL

emphysema PERSON

4 CARDINAL

5 CARDINAL

4 CARDINAL

8) CARDINAL

first ORDINAL

emphysema PERSON

secondly ORDINAL

emphysema PERSON

Delhi GPE

'Delhi GPE

Delhi GPE

Indian NORP

Delhi GPE

Indian NORP

Delhi GPE

'Delhi GPE

emphysema PERSON

one CARDINAL

BioFlex ORG

firstly ORDINAL

secondly ORDINAL

the University of Arkansas ORG

Villanova University ORG

Marguette University ORG

more than 500 CARDINAL

U.S. GPE

Canadian NORP

Scot Burton PERSON

the Sam M. Walton College of Business ORG

two CARDINAL

EFCC ORG

Theodore Orji PERSON

N521 billion MONEY

ICE Detainer Issued for Suspected Wine Country Arsonist ORG

Sonoma Jail GPE

ASUU ORG

Another Violent Leftist Arrested ORG

US Capital Protest ORG

Democrat NORP

34-Year-Old DATE

New York GPE

Biden Motorcade PERSON

Jay Leno PERSON

Republican NORP

Trump Supporter PERSON

71 DATE

FAKED DATE

Steven Seagal PERSON

Putin PERSON

Russia GPE

Far East LOC

Biden PERSON

Black Delaware State FAC

200,000 MONEY

**DEMOCRATS NORP** 

B.C. GPE

ALS ORG

vears DATE

24-hour TIME

Fox News ORG

Jeanine Show PERSON

over 200,000 CARDINAL

NHS ORG

2010 DATE

£330 million MONEY

CO2 CARDINAL

Senate ORG

House ORG

over 4 million CARDINAL

Minneapolis GPE

George Floyd PERSON

Derek Chauvin PERSON

1984 DATE

the New York Times ORG

Trump ORG

Conservatives NORP

MSF ORG

18 CARDINAL

first ORDINAL

Britain GPE

36 MONEY

Israeli NORP

first ORDINAL

vear DATE

Dan Aridor PERSON

Accelerated Evolution Biotechnologies Ltd. ORG

2000 DATE

ITEK GPE

SoAP ORG

day one DATE

a few weeks DATE

Aridor PERSON

Ilan Moran PERSON

Ilan Moran PERSON

an estimated 18.1 million MONEY

each year DATE

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the International Agency for Research on Cancer ORG
sixth ORDINAL
second ORDINAL
second ORDINAL
Aridor ORG
Ilan Morad PERSON
SoAP ORG
2018 DATE
the Nobel Prize WORK OF ART
two CARDINAL
Morad GPE
first ORDINAL
Morad GPE
Morad GPE
Hanan Itzhaki PERSON
Morad GPE
at least three CARDINAL
Morad GPE
Morad GPE
three CARDINAL
three CARDINAL
Morad GPE
Morad PERSON
Morad PERSON
12 CARDINAL
Morad PERSON
one CARDINAL
two CARDINAL
Morad GPE
Today DATE
two CARDINAL
Morad GPE
only a few weeks DATE
Aridor GPE
Morad PERSON
first ORDINAL
a few years DATE
Colin Powell PERSON
Michael Flynn PERSON
Biden PERSON
Friends Busted Partying and Singing Without Masks or Social Distancing
ORG
One CARDINAL
Tetyana Obukhanych ORG
Ph.D. WORK_OF_ART
the Rockefeller University ORG
New York GPE
Harvard ORG
British Columbia GPE
Ouebec GPE
```

2011 DATE

48% PERCENT

PhD WORK OF ART

one CARDINAL

two CARDINAL

a few years ago DATE

American NORP

first ORDINAL

two CARDINAL

11 DATE

Ukraine GPE

PhD WORK OF ART

One CARDINAL

Hurricane Patricia EVENT

50,000 MONEY

Capitol Shooting Exposed ORG

Hollywood GPE

ASUU ORG

JK Rowling Confirms Stance Against Transgender Women ORG

78 CARDINAL

Epstein Isle LOC

one hour TIME

Biden PERSON

the World Health Organization ORG

PCR ORG

WHO ORG

today DATE

COVID ORG

WHO ORG

COVID ORG

March 2020 DATE

WHO ORG

COVID ORG

COVID ORG

NHS ORG

the United States GPE

Rick Simpson PERSON

marijuana PERSON

Cannabis PERSON

Marijuana PERSON

Meghan Markle PERSON

America GPE

first ORDINAL

MSM ORG

Italy GPE

Nazi NORP

pro-Russian NORP

the War on Drugs EVENT

2014 DATE

Breaks Heat Record EVENT

Antifa Provocateur PERSON

Earth LOC

400th FAC

British Government Study Confirms Covid-19 Vaccine Risk: Infections

**ORG** 

Fortnight GPE

January DATE

80 CARDINAL

Pemberley House ORG

Basingstoke GPE

England GPE

a third CARDINAL

24 CARDINAL

Germany GPE

British NORP

Gibraltar GPE

Pfizer ORG

16th January DATE

33 thousand CARDINAL

2761 CARDINAL

9 CARDINAL

December DATE

almost a hundred CARDINAL

The United Arab Emirates GPE

Chinese NORP

Sinopharm ORG

Pfizer ORG

December DATE

Amer Sharif PERSON

the Covid-19 Command & Control Centre ORG

Dubai GPE

UAE ORG

first ORDINAL

100% PERCENT

10th February DATE

daily DATE

18 CARDINAL

first ORDINAL

last spring DATE

Israel GPE

19th December DATE

New York Times ORG

Alex Berenson PERSON

Israelis NORP

January DATE

Palestine GPE

December DATE

Palestinians NORP

Israeli NORP

Hervé Seligmann PERSON

Aix-Marseilles University ORG

about 40 CARDINAL

the three weeks DATE

first ORDINAL

second ORDINAL

Israel GPE

Pfizer ORG

the first week DATE

Raine PERSON

UK GPE

MHRA ORG

two CARDINAL

first ORDINAL

The Public Health England ORG

Matt Hancock PERSON

80% PERCENT

48% PERCENT

first ORDINAL

Pfizer ORG

Astra ORG

Zeneca NORP

Mary Ramsay PERSON

PHE ORG

hundreds CARDINAL

British NORP

the Bill & Melinda Gates Foundation LAW

Covid PERSON

Covid PERSON

Mike Pence PERSON

Philadelphia GPE

Penn State ORG

Craig Meyers PERSON

NewsNation ORG

WTAJ ORG

the Journal of Medical Virology ORG

the Penn State College of Medicine ORG

Listerene PERSON

at least 30 seconds TIME

Biden PERSON

Friends Busted Partying and Singing Without Masks or Social Distancing

**ORG** 

Ireland GPE

one CARDINAL

two CARDINAL

Alaska GPE

Permafrost Is Thawing WORK\_OF\_ART

Nasa ORG

ICE Detainer Issued for Suspected Wine Country Arsonist ORG

Sonoma Jail GPE

Colorado Rockies Baseball Team ORG

Erie GPE

To Throwing Away 'Over A Hundred' Pro-Trump Ballots On Instagram

WORK OF ART

Son PERSON

Daughter PERSON

Exclusive: The TCF Center Election Fraud — Newly Discovered Video

Shows Late Night Deliveries WORK OF ART

Tens of Thousands CARDINAL

8 Hours TIME

Gingrich Tragedy - PERSON

Obama PERSON

November 14, 2014 DATE

Olympics EVENT

Rio Games ORG

Coronavirus Contains HIV Insertions ORG

Global Ocean Circulation ORG

Rick Perry PERSON

111 CARDINAL

NHS ORG

N.J. GPE

500 MONEY

Supreme Court ORG

Eugenie PERSON

ninth ORDINAL

Killers ORG

Sentencing Council ORG

2040 DATE

Democrat NORP

Anti-Trump Postal Workers ORG

Garbage ORG

Pfizer PERSON

CLARKSBURG ORG

Jared Henry ORG

Maddie Mansell PERSON

Buckhannon Pike PERSON

Nutter Fort GPE

Mansell PERSON

Henry PERSON

Mansell PERSON

Mansell PERSON

Nikki Bowmar PERSON

the U.S. Small Business Administration ORG

Bowmar PERSON

approximately 114,391 CARDINAL

West Virginia GPE

99 percent PERCENT

280,213 CARDINAL

nearly 50 percent PERCENT

Bowmar PERSON

Mansell PERSON

first ORDINAL

Christ GPE

George "Gib" Brown PERSON

The James and Law Co. ORG

West Virginia GPE

nearly 30 CARDINAL

Clarksburg GPE

two CARDINAL

1903 DATE

Brown PERSON

James and Law ORG

Brown PERSON

the years DATE

Chamber ORG

the Robinson Grand Performing Arts Center ORG

Clarksburg GPE

Clarksburg GPE

years DATE

Brown PERSON

Dan Taylor PERSON

the West Virginia Community Development Hub's Energizing

Entrepreneurial Communities ORG

the last of three years DATE

four CARDINAL

Grafton ORG

Taylor PERSON

Taylor PERSON

Stephen Mann PERSON

Lewis County GPE

December DATE

Weston GPE

Mann ORG

two CARDINAL

one CARDINAL

401k PRODUCT

Mann PERSON

Manns PERSON

Lewis County GPE

Mann PERSON

Lewis County GPE

West Virginia GPE

Bowmar PERSON

Mann ORG

The Energizing Entrepreneurial Communities ORG

Grafton ORG

Taylor PERSON

All Aboard Grafton FAC

Taylor PERSON

Taylor County PERSON

first ORDINAL

February DATE

Patricia Henderson PERSON

the Taylor County Development Authority ORG

the Taylor County Commission ORG

Taylor County GPE

Weston GPE

Fairs and Festivals Committee ORG

Julia Spelsberg PERSON

Spelsberg PERSON

Weston GPE

North Central West Virginia LOC

The Small Business Administration ORG

Bowmar PERSON

USDA ORG

the Tax Cut FAC

Taylor PERSON

Obamacare NORP

Months DATE

the Texas House ORG

Texas Scorecard ORG

Empower Texans NORP

one CARDINAL

First Amendment LAW

January DATE

two CARDINAL

Texas Scorecard ORG

Brandon Waltens PERSON

Destin Sensky PERSON

Legislature ORG

Senate ORG

House ORG

two CARDINAL

House Administration Chair ORG

Charlie Geren PERSON

Fort Worth GPE

Republican NORP

Empower Texans NORP

PAC ORG

January DATE

Texas Scorecard ORG

PAC ORG

Geren PERSON

Geren PERSON

Empower Texans NORP

PAC ORG

Geren PERSON

Dan Patrick PERSON

the Texas Senate ORG

more than \$850,000 MONEY

the last five years DATE

this week DATE

Austin GPE

Geren PERSON

Geren PERSON

two CARDINAL

the Texas House ORG

Wednesday DATE

morning TIME

Geren PERSON

Empower Texans NORP

weekday DATE

weekly DATE

A-F PRODUCT

the Texas House ORG

Waltens PERSON

Sensky PERSON

second ORDINAL

House ORG

Geren PERSON

Empower Texans NORP

2017 DATE

Geren PERSON

Cary Cheshire Waltens PERSON

Sensky ORG

the Texas Senate ORG

the Texas House ORG

Illinois General Assembly ORG

Florida GPE

Bloomberg PERSON

Caruthers California: Video LOC

Vladimir Putin PERSON

Europe LOC

Vigil PERSON

13-year-old DATE

The New York Times ORG

TRUMP ORG

Obama GPE

Reagan PERSON

Same Point ORG

Presidencies FAC

Antifa PERSON

Trump ORG

Scott PERSON

Government ORG

Kirk Douglas Dead PERSON

4 Days DATE

Apocalyptic Sea-Level Rise ORG

15 CARDINAL

Nigerian House of Reps ORG

Coronavirus PERSON

Gossip PERSON

Nigeria GPE

Fox News Wins Pulitzer Prize for Reporting During Crisis — Daily Worl

Biden Calls PERSON

Done Meeting PERSON

NIPOST ORG

Osogbo GPE

Biden Motorcade PERSON

Childhood ORG

40% PERCENT

four years DATE

Jay Leno PERSON

Republican NORP

Trump Supporter PERSON

71 DATE

Democrats NORP

\$2.2 Billion MONEY

Social Security ORG

Cuomo's Budget ORG

U.S. GPE

Jeff Merkley PERSON

Oregon GPE

Joe Biden's Dept PERSON

Energy Blocked GPE

Texas GPE

Increasing Power Ahead of Killer Storm ORG

Biden Plans PERSON

Move Social Security Eligibility ORG

75 CARDINAL

1,500 MONEY

Next Week DATE

More Than 10,000 CARDINAL

Cast Ballots PERSON

Michigan GPE

UK GPE

4% PERCENT

Ucas GPE

NIPOST ORG

Osogbo GPE

Tory ORG

75 CARDINAL

the next 16 years DATE

Fani-Kayode PERSON

Abba Kyari's PERSON

Wolf PERSON

Britain GPE

Colorado Rockies Baseball Team ORG

Ariana Grande No Longer ORG

Japanese NORP

Obama PERSON

Cameron PERSON

the United Kingdom GPE

Michigan GPE

Travis Rowley PERSON

Democrats NORP

CO2 CARDINAL

A decade DATE

Scottish National Party ORG

Feds Handcuff 'Irrational' House PERSON

Raiding Her Office ORG

San Diego GPE

Democrat NORP

Peters PERSON

Trump ORG

Mexico GPE

MSF ORG

18 CARDINAL

first ORDINAL

BOMBSHELL ORG

440% PERCENT

the coming flu season DATE

Sarah Parker PERSON

Supreme Court ORG

DeSantis Tells PERSON

Go Fuck Yourself" WORK\_OF\_ART

CDC Inflated COVID Numbers ORG

Russia GPE

Trump ORG

2020 DATE

FG ORG

EU ORG

Kwik PERSON

Launches Kwik Strip Gentlemen's PERSON

2030 DATE

London GPE

Apollo ORG

Alien City GPE

Burnam Statement PERSON

GOP ORG

Vote Rejecting Medicaid Expansion ORG

NHS ORG

US GPE

One Hour TIME

Biden PERSON

Arizona Senate ORG

Releases Evidence PERSON

Democrat Fundraising Machine 'ActBlue Texas ORG

Caught Paying Out Small Sums WORK OF ART

Hundreds of Individuals for Unknown Purposes - David Harris Jr ORG

Gohmert Responds PERSON

May 2018 DATE

fourth ORDINAL

the month of May DATE

NOAA ORG

1880 DATE

March-May DATE

fourth ORDINAL

monthly DATE

NOAA ORG

National Centers for Environmental Information ORG

NOAA ORG

May 2018 DATE

May DATE

1.44 CARDINAL

F PRODUCT

the 20th century DATE

58.6 CARDINAL

F PRODUCT

fourth ORDINAL

May DATE

1880-2018 DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.58 CARDINAL

May 2018 DATE

the 42nd consecutive May DATE

the 401st consecutive month DATE

the 20th century DATE

North America LOC

Europe LOC

Asia LOC

Atlantic LOC

Pacific LOC

Indian Oceans NORP

May DATE

Canada GPE

Atlantic Ocean LOC

Greenland GPE

May DATE

2.05 CARDINAL

the 20th century DATE

52.0 CARDINAL

F PRODUCT

May DATE

2011 DATE

2013 DATE

seventh ORDINAL

May DATE

139-year DATE

May 2018 DATE

U.S. GPE

Europe LOC

5.4 CARDINAL

Canada GPE

Russia GPE

5.4 CARDINAL

Europe LOC

May DATE

continental ORG

1910 DATE

4.97 CARDINAL

2003 DATE

1.66 CARDINAL

F PRODUCT

May 2018 DATE

first ORDINAL

May DATE

3.6 CARDINAL

F PRODUCT

2.0 CARDINAL

European NORP

May DATE

Asia LOC

May DATE

2009 DATE

May DATE

1.19 CARDINAL

the 20th century DATE

monthly DATE

61.3 CARDINAL

fourth ORDINAL

May DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.39 CARDINAL

May 2018 DATE

May DATE

Arctic LOC

second ORDINAL

40-year DATE

420,000 square miles QUANTITY

8.1 percent PERCENT

1981-2010 DATE

the National Snow and Ice Data Center(link ORG

NOAA ORG

NASA ORG

May Arctic LOC

2016 DATE

Arctic LOC

Bering GPE

Chukchi Seas PERSON

May DATE

330,000 square miles QUANTITY

8.6 percent PERCENT

1981-2010 DATE

third ORDINAL

May DATE

Antarctic sea LOC

May DATE

the Weddell Sea LOC

Seasonal DATE

March-May 2018 DATE

March-May DATE

1.48 CARDINAL

the 20th century DATE

56.7 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

1880-2018 DATE

0.45 CARDINAL

F cooler PRODUCT

2016 DATE

March-May DATE

2014 DATE

three-month DATE

Europe LOC

the Middle East LOC

Africa LOC

China GPE

Mongolia GPE

South America LOC

Atlantic LOC

Indian NORP

Pacific Oceans NORP

March-May 2018 DATE

March-May DATE

2.34 CARDINAL

the 20th century DATE

46.4 CARDINAL

F PRODUCT

fifth ORDINAL

March-May DATE

139-year DATE

Five CARDINAL

March-May 2018 DATE

nine CARDINAL

continental ORG

1910 DATE

Europe LOC

South America LOC

third ORDINAL

March-May DATE

March-May DATE

1.15 CARDINAL

the 20th century DATE

61.0 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

Year-to-Date DATE

January-May 2018 DATE

1.39 CARDINAL

the 20th century DATE

55.5 CARDINAL

F PRODUCT

2010 DATE

fourth ORDINAL

January-May DATE

139-year DATE

2018 year DATE

0.59 CARDINAL

2016 DATE

2.18 CARDINAL

the 20th century DATE

42.8 CARDINAL

F PRODUCT

2002 DATE

fifth ORDINAL

January-May DATE

1.08 CARDINAL

the 20th century DATE

60.8 CARDINAL

fifth ORDINAL

January-May DATE

1880-2018 DATE

Democrats NORP

Americans NORP

9 CARDINAL

Capitol Shooting Exposed ORG

Hollywood GPE

Up to a third CARDINAL

Iran GPE

Erbil ORG

Iraq GPE

Pittsburg Area Antifa LOC

Tweets Warning PERSON

Trump WORK OF ART

Sunday DATE

noon TIME

Joe Biden PERSON

Government ORG

Erie GPE

To Throwing Away 'Over A Hundred' Pro-Trump Ballots On Instagram

WORK\_OF\_ART

2017 DATE

Paid Liberal Trolls of America ORG

6 hours TIME

30 MONEY

40 MONEY

70 cents MONEY

1 MONEY

Soros PERSON

Joe Barron PERSON

George Soros PERSON

500 MONEY

Art Tubolls PERSON

163 CARDINAL

Soros PERSON

Deluth GPE

Charlie Kirk PERSON

Turning Point PERSON

The Fonz" Winkler, Dies At Age 74 After Secret Battle With Guillain-

Barré Syndrome GBS WORK OF ART

CDC ORG

the American Academy of Pediatrics ORG

Andrew Zimmerman PERSON

Kennedy Krieger Institute ORG

Johns Hopkins University ORG

third ORDINAL

years DATE

Zimmerman PERSON

Zimmerman PERSON

more than five thousand CARDINAL

2009 DATE

the late 1990s DATE

the Kennedy Krieger Institute ORG

Baltimore GPE

Zimmerman PERSON

Jon Poling PERSON

2000 DATE

nineteen-month-old DATE

Hannah PERSON

Hannah PERSON

Zimmerman PERSON

Richard Kelley PERSON

Kennedy Krieger's PERSON

Kennedy Krieger PERSON

Hannah PERSON

Georgetown GPE

MD GPE

Teri PERSON

Hannah PERSON

the next few years DATE

Zimmerman PERSON

2008 DATE

Hannah Poling PERSON

Zimmerman PERSON

Zimmerman PERSON

Zimmerman PERSON

Hannah PERSON

CDC ORG

the Department of Justice ORG

twenty-million-dollar MONEY

Polings FAC

2010 DATE

Hannah PERSON

Hannah Poling PERSON

Zimmerman PERSON

Kelley PERSON

the Department of Justice ORG

Drs PERSON

Zimmerman PERSON

Kelley PERSON

first ORDINAL

more than thirty years DATE

Tennessee GPE

Yates Hazlehurst PERSON

one CARDINAL

three CARDINAL

2009 DATE

only a year DATE

Poling ORG

Zimmerman.3 Drs PERSON

Zimmerman PERSON

Kelley PERSON

Hazlehurst NORP

Yates GPE

2009 DATE

the Omnibus Autism Proceeding ORG

Zimmerman PERSON

more than five thousand CARDINAL

2017 DATE

Drs PERSON

Kelley PERSON

Zimmerman PERSON

Yates GPE

The "Vaccine Court ORG

the National Childhood Vaccine Injury Act ORG

1986 DATE

the United States Court of Federal Claims ORG

American NORP

the Department of Health and Human Services ORG

HHS ORG

Department of Justice ORG

1989 DATE

more than \$3.8 billion MONEY

Americans NORP

1998 DATE

twelve hundred CARDINAL

eighteen thousand CARDINAL

DTP ORG

MMR ORG

second ORDINAL

approximately 34 percent PERCENT

2017 DATE

just over \$282 million MONEY

Rolf Hazlehurst PERSON

Tennessee GPE

Yates PERSON

the US Congress ORG

2013 DATE

Rolf Hazlehurst ORG

Vaccine PERSON

Discovery PRODUCT

American NORP

ENCYCLICAL LETTER LAUDATO SI' OF THE HOLY FATHER FRANCIS ON CARE FOR

OUR COMMON HOME ORG

Will Austin PERSON

\$1 Million MONEY

GEORGIA GPE

95-Degree Days DATE

NHS ORG

30% PERCENT

Paul Bettencourt PERSON

BRAZIL GPE

AstraZeneca ORG

Oxford University ORG

Anvisa ORG

one CARDINAL

three CARDINAL

Brazil GPE

Oxford University ORG

the UK Government ORG

100 million CARDINAL

US GPE

September 6 DATE

India GPE

South Africa GPE

AstraZeneca ORG

Europe LOC

Inclusive Vaccine Alliance ORG

400 million CARDINAL

European NORP

UK GPE

30 million CARDINAL

BioNtech/Pfizer ORG

60 million CARDINAL

Valneva - in PERSON

six CARDINAL

UK GPE

Patrick Vallance PERSON

next spring DATE

the Commons and Lords National Security Strategy Committee ORG

AstraZeneca ORG

the turn of the year DATE

UK GPE

Vaccine Taskforce FAC

Kate Bingham PERSON

UK GPE

four CARDINAL

six CARDINAL

three CARDINAL

one CARDINAL

two CARDINAL

Covid PERSON

26,688 CARDINAL

UK GPE

789,229 CARDINAL

191 CARDINAL

today DATE

Earth LOC

Antifa PERSON

Trump ORG

this week DATE

Scottish National Party ORG

Democrats NORP

Americans NORP

More Than 40% PERCENT

U.S. GPE

Australia GPE

Great Barrier Reef WORK OF ART

at least 40,000 CARDINAL

UK GPE

Earth LOC

400th FAC

NHS ORG

1bn ORDINAL

Brexit PERSON

Britons GPE

EU ORG

Trump ORG

Lin Wood Puts PERSON

Gov Brian Kemp PERSON

JUSTICE ORG

EU ORG

Arctic LOC

first ORDINAL

Two CARDINAL

Georgia GPE

\$2.2 Billion MONEY

Social Security ORG

Cuomo's Budget ORG

Democrats NORP

Bill To 'Euthanize Seniors' PERSON

Childhood ORG

40% PERCENT

four years DATE

Olympics EVENT

Rio Games ORG

Obama's Daughters Caught ORG

Atlanta GPE

JOINS ORG

GEORGIA GPE

**NEVADA GPE** 

TRUMP GPE

Robert F. Kennedy Jr PERSON

Facebook Fact Checkers ORG

four million CARDINAL

a year DATE

Barack Obama PERSON

2000 years DATE

the Medieval Warm Period FAC

Little Ice Age LOC

JoNova ORG

Obama PERSON

Cameron PERSON

the United Kingdom GPE

WHO ORG

Vladimir Putin PERSON

Europe LOC

Up to 90% PERCENT

Covid PERSON

CDC ORG

1600% PERCENT

Josh Mandel PERSON

USCIS ORG

2017 DATE

Arctic LOC

Winter Sea Ice Drops LOC

UK GPE

Oregon GPE

Hundreds CARDINAL

Republican NORP

GOP Voters ORG

Candace Owens PERSON

Arctic LOC

first ORDINAL

Trump ORG

Statue PERSON

Trump ORG

Vladimir Putin PERSON

FDA Shocking Study: Cells Used ORG

ENCYCLICAL LETTER LAUDATO SI' OF THE HOLY FATHER FRANCIS ON CARE FOR

OUR COMMON HOME ORG

CDC ORG

COVID-19 ORG

U.S. GPE

CDC ORG

between March and December 2020 DATE

1.5 percent PERCENT

two months DATE

0.5 percent PERCENT

the first 20 days DATE

less than 2 percent PERCENT

100 days DATE

CDC ORG

11-year-olds ORDINAL

200,000 MONEY

America GPE

Zamfara PERSON

Charlie Crist PERSON

Nan Rich ORG

Washington State GPE

2014 DATE

Breaks Heat Record EVENT

June DATE

Earth LOC

European NORP

British NORP

100 years ago DATE

Miami GPE

Caruthers California: Video LOC

One CARDINAL

Tilapia ORG

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Tilapia ORG
Wake Forest University School of Medicine ORG
Tilapia PERSON
Tilapia ORG
Tilapia ORG
one CARDINAL
Tilapia PERSON
Tilapia ORG
Tilapia PERSON
Tilapia PERSON
Tilapia GPE
China GPE
October DATE
the White House Arresting Congress FAC
Bible WORK OF ART
Buy Isa ORG
500,000 CARDINAL
first ORDINAL
CDC ORG
COVID-19 ORG
U.S. GPE
CDC ORG
between March and December 2020 DATE
1.5 percent PERCENT
two months DATE
0.5 percent PERCENT
the first 20 days DATE
less than 2 percent PERCENT
100 days DATE
CDC ORG
Kwik PERSON
Launches Kwik Strip Gentlemen's PERSON
first ORDINAL
This Labor Day DATE
Rand Paul PERSON
Scott Walker PERSON
$100 million MONEY
US GPE
ISIS ORG
Syrian NORP
Arctic LOC
Antarctic LOC
25 years DATE
the U.S. Centers for Disease Control and Prevention ORG
CDC ORG
the Vaccine Adverse Event Reporting System ORG
more than 10 years DATE
at least 108 CARDINAL
today DATE
today DATE
```

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first ORDINAL
U.S. GPE
2003 DATE
the Associate Press ORG
Anne Schuchat PERSON
CDC ORG
National Center for Immunization and Respiratory Diseases ORG
VAERS ORG
at least 108 CARDINAL
2003 DATE
96 CARDINAL
MMR ORG
first ORDINAL
1963 DATE
U.S. GPE
the years DATE
first ORDINAL
1963 DATE
the United States GPE
approximately 98% PERCENT
the International Medical Council on Vaccination ORG
first ORDINAL
IMCV PERSON
One CARDINAL
COVID-19 ORG
COVID ORG
November 20th DATE
almost 10 million CARDINAL
China GPE
300 CARDINAL
Coronavirus GPE
1.174 CARDINAL
ZERO CARDINAL
zero CARDINAL
Coronavirus GPE
10,000,000 CARDINAL
Wuhan GPE
between January 23 and April 8, 2020 DATE
May 14 and June 1, 2020 DATE
Wuhan GPE
six years DATE
9,899,828 CARDINAL
92.9% PERCENT
300 CARDINAL
0.303/10,000 CARDINAL
95% PERCENT
CI ORG
```

1,174 CARDINAL 107 CARDINAL 34,424 CARDINAL COVID-19 PRODUCT

0.31% PERCENT

95% PERCENT

0.423-0.574% PERCENT

Wuhan GPE

five to eight weeks DATE

10 million CARDINAL

first ORDINAL

10 million CARDINAL

World Health Organization ORG

COVID-19 ORG

November DATE

the World Health Organization ORG

COVID-19 ORG

June of 2020 DATE

Maria Van Kerkhove PERSON

the World Health Organization's ORG

Bloomberg Markets ORG

Finance ORG

847 CARDINAL

WHO ORG

June 2020 DATE

Kerkhove PERSON

2020 CARDINAL

COVID-19 ORG

Capitol FAC

Legislature ORG

Australian NORP

Democrats NORP

Trump ORG

McCain ORG

Marijuana PERSON

Michigan GPE

Jon Evans PERSON

Amazon ORG

Tim Ball: PERSON

Will Austin PERSON

2000 years DATE

the Medieval Warm Period FAC

Little Ice Age LOC

JoNova ORG

WADA GPE

Moscow GPE

IOC ORG

Olympics EVENT

Amazon ORG

Greenland GPE

annual DATE

China GPE

Canada GPE

The Ron Paul Institute for Peace and Prosperity: The Coronavirus Hoax **ORG** Brexit Britain PERSON EU ORG SURGES ORG El Niño ORG 2015 DATE Hillary Caught PERSON Russian NORP Syria GPE 2017 DATE Reuters ORG 2,500 MONEY Banned From Driving For Smoking ORG Canada GPE Conservatives NORP Ilhan Omar Tells PERSON Minnesota LOC Earth LOC Trump ORG Statue PERSON Democrats NORP Netizens ORG ngayon ORG China GPE Scott Walker PERSON \$100 million MONEY **Brexit PERSON** Britons GPE EU ORG billions CARDINAL Mueller ORG Russia GPE more than \$25 million MONEY 50,000 MONEY Belgian NORP Fremault PERSON Brussels GPE 5 CARDINAL Celine Fremault PERSON the Brussels-Capital Region Responsible For Housing, Quality Of Life, Environment And Energy ORG 5 CARDINAL Brussels GPE Fremault PERSON 5 CARDINAL Belgian NORP 9 CARDINAL 95 mW/ QUANTITY m2 PERSON

1 mW/m2 QUANTITY

thousands CARDINAL

Brussels GPE

EU ORG

one CARDINAL

2 CARDINAL

5 CARDINAL

EU ORG

EU ORG

Celine Fremault PERSON

North America LOC

Celine Fremault PERSON

Brussel GPE

5 CARDINAL

6 minute TIME

UN ORG

5 CARDINAL

zero CARDINAL

Tom Wheeler PERSON

FCC Chair ORG

5 CARDINAL

5 CARDINAL

5 CARDINAL

5 CARDINAL

US GPE

Hundreds CARDINAL

Martin Pall PERSON

5 CARDINAL

Earth LOC

5 CARDINAL

AI ORG

5 CARDINAL

Brussels GPE

first ORDINAL

5 CARDINAL

dozens CARDINAL

5 CARDINAL

21 CARDINAL

**USA GPE** 

March 13, 2019 DATE

Portland GPE

FCC ORG

EMF Safety Network ORG

WhatIs5G.info ORG

5 CARDINAL

Fox News Wins Pulitzer Prize for Reporting During Crisis — Daily Worl ORG

N.J. GPE

500 MONEY

Supreme Court ORG

Government ORG

More Than 40% PERCENT

U.S. GPE

EU ORG

Tammy Baldwin PERSON

Wisconsin GPE

**Election Year DATE** 

Buhari To Resign PERSON

May 2023 DATE

Lai Mohammed PERSON

Apollo ORG

Alien City GPE

2015 DATE

The New York Times ORG

Fani-Kayode PERSON

Abba Kyari's PERSON

Burnam Statement PERSON

GOP ORG

Vote Rejecting Medicaid Expansion ORG

Muslim NORP

Canada GPE

at least 13 CARDINAL

COVID-19 ORG

the United States of America GPE

the past few weeks DATE

COVID-19 ORG

COVID-19 ORG

Donald Trump PERSON

Canada GPE

800 CARDINAL

Cannabis PRODUCT

COVID-19 ORG

13 CARDINAL

ACE2 PRODUCT

C. NORP

TMPRSS2 PRODUCT

COVID-19 PERSON

13 CARDINAL

Canada's CARDINAL

Trump ORG

Harvard Medical School ORG

Mandeep Mehra PERSON

more than 96,000 CARDINAL

15,000 CARDINAL

the Washington Post ORG

34 percent PERCENT

137 percent PERCENT

Trump ORG

45 percent PERCENT

411 percent PERCENT

37 percent PERCENT

256 percent PERCENT

37 percent PERCENT

301 percent PERCENT

Donald Trump PERSON

US GPE

COVID-19 ORG

Plaquenil NORP

COVID-19 PERSON

COVID-19 ORG

zero-hours TIME

Britain GPE

Windrush PERSON

SNUBBED ORG

week DATE

Commonwealth ORG

FAKED DATE

Trudeau Bought PERSON

Arctic LOC

Antarctic LOC

Killers ORG

Sentencing Council ORG

Russian NORP

Syria GPE

2017 DATE

Reuters ORG

CO2 CARDINAL

Biden PERSON

Russia GPE

SolarWinds ORG

500 MONEY

**Brexit PERSON** 

Netizens ORG

ngayon ORG

China GPE

Hayes PERSON

German NORP

COVID-19 ORG

30 Days DATE

Pakistani NORP

Meesha Shafi PERSON

3 years DATE

Ali Zafar PERSON

Clean Energy and Pollution Reduction Act ORG

2015 DATE

over 4 million CARDINAL

Erpenbach PERSON

Walker PERSON

Health Insurance Exchanges ORG

75 CARDINAL

MPH ORG

six months DATE

two-year-old DATE

months DATE

COVID ORG

one CARDINAL

about 6-8 weeks CARDINAL

October 23 DATE

MIT ORG

the University of California San Francisco ORG

Decrease in Hospitalizations for COVID-19 WORK\_OF\_ART

Mask Mandates PERSON

1083 DATE

U.S. GPE

November 4 DATE

months DATE

COVID ORG

one CARDINAL

about 6-8 weeks CARDINAL

October 23 DATE

MIT ORG

the University of California San Francisco ORG

Decrease in Hospitalizations for COVID-19 WORK OF ART

Mask Mandates PERSON

1083 DATE

U.S. GPE

November 4 DATE

second ORDINAL

third ORDINAL

Danish NORP

months DATE

Last week DATE

the Naval Medical Research Center ORG

Maryland GPE

Mount Sinai Hospital FAC

New York GPE

the New England Journal of Medicine ORG

U.S. Marine Corps ORG

Parris Island GPE

over 3,000 CARDINAL

two CARDINAL

day 1 DATE

day 14 DATE

2.8% PERCENT

1.7% PERCENT

at least 6 feet QUANTITY

daily DATE

Six CARDINAL

8-hour TIME

Fauci PERSON

Marine NORP

Army ORG

Fort Benning FAC

90% PERCENT

Trump ORG

thousands CARDINAL

Wisconsin GPE

America GPE

Zamfara PERSON

99.96% PERCENT

two CARDINAL

Pfizer ORG

Covid PERSON

Israel GPE

Miles PERSON

COVID-19 ORG

COVID ORG

November 20th DATE

almost 10 million CARDINAL

China GPE

300 CARDINAL

Coronavirus GPE

1,174 CARDINAL

ZERO CARDINAL

zero CARDINAL

Coronavirus GPE

10,000,000 CARDINAL

Wuhan GPE

between January 23 and April 8, 2020 DATE

May 14 and June 1, 2020 DATE

Wuhan GPE

six years DATE

9,899,828 CARDINAL

92.9% PERCENT

300 CARDINAL

0.303/10,000 CARDINAL

95% PERCENT

CI ORG

1.174 CARDINAL

107 CARDINAL

34,424 CARDINAL

COVID-19 PRODUCT

0.31% PERCENT

95% PERCENT

0.423-0.574% PERCENT

Wuhan GPE

five to eight weeks DATE

10 million CARDINAL

first ORDINAL

10 million CARDINAL

World Health Organization ORG

COVID-19 ORG

November DATE

the World Health Organization ORG

COVID-19 ORG

June of 2020 DATE

Maria Van Kerkhove PERSON

the World Health Organization's ORG

Bloomberg Markets ORG

Finance ORG

847 CARDINAL

WHO ORG

June 2020 DATE

Kerkhove PERSON

2020 CARDINAL

COVID-19 ORG

21 Million CARDINAL

China GPE

Coronavirus Outbreak GPE

Olympics EVENT

Team GB PERSON

China GPE

Rio Olympics EVENT

Steven Seagal PERSON

Putin PERSON

Russia GPE

Far East LOC

HPV ORG

Joe Biden Tells PERSON

Joe Biden PERSON

DISGRACEFUL ORG

EU ORG

British NORP

140MILLION MONEY

Antifa Provocateur PERSON

Pope PERSON

80 CARDINAL

Child Trafficking, Fraud — Conservative WORK OF ART

at least 40,000 CARDINAL

UK GPE

3,000 MONEY

Canada GPE

Today DATE

Marine Corps ORG

Justine Greening PERSON

Tom Hanks PERSON

65 DATE

Passes Away PERSON

May 2018 DATE

fourth ORDINAL

the month of May DATE NOAA ORG 1880 DATE March-May DATE fourth ORDINAL monthly DATE NOAA ORG National Centers for Environmental Information ORG NOAA ORG May 2018 DATE May DATE 1.44 CARDINAL F PRODUCT the 20th century DATE 58.6 CARDINAL F PRODUCT fourth ORDINAL May DATE 1880-2018 DATE years 2014-2018 DATE five CARDINAL 2016 CARDINAL May DATE 1.58 CARDINAL May 2018 DATE the 42nd consecutive May DATE the 401st consecutive month DATE the 20th century DATE North America LOC Europe LOC Asia LOC Atlantic LOC Pacific LOC Indian Oceans NORP May DATE Canada GPE Atlantic Ocean LOC Greenland GPE May DATE 2.05 CARDINAL the 20th century DATE 52.0 CARDINAL

the 20th century DATE 52.0 CARDINAL F PRODUCT May DATE 2011 DATE 2013 DATE seventh ORDINAL May DATE 139-year DATE May 2018 DATE

U.S. GPE

Europe LOC

5.4 CARDINAL

Canada GPE

Russia GPE

5.4 CARDINAL

Europe LOC

May DATE

continental ORG

1910 DATE

4.97 CARDINAL

2003 DATE

1.66 CARDINAL

F PRODUCT

May 2018 DATE

first ORDINAL

May DATE

3.6 CARDINAL

F PRODUCT

2.0 CARDINAL

European NORP

May DATE

Asia LOC

May DATE

2009 DATE

May DATE

1.19 CARDINAL

the 20th century DATE

monthly DATE

61.3 CARDINAL

fourth ORDINAL

May DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.39 CARDINAL

May 2018 DATE

May DATE

Arctic LOC

second ORDINAL

40-year DATE

420,000 square miles QUANTITY

8.1 percent PERCENT

1981-2010 DATE

the National Snow and Ice Data Center(link ORG

NOAA ORG

NASA ORG

May Arctic LOC

2016 DATE

Arctic LOC

Bering GPE

Chukchi Seas PERSON

May DATE

330,000 square miles QUANTITY

8.6 percent PERCENT

1981-2010 DATE

third ORDINAL

May DATE

Antarctic sea LOC

May DATE

the Weddell Sea LOC

Seasonal DATE

March-May 2018 DATE

March-May DATE

1.48 CARDINAL

the 20th century DATE

56.7 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

1880-2018 DATE

0.45 CARDINAL

F cooler PRODUCT

2016 DATE

March-May DATE

2014 DATE

three-month DATE

Europe LOC

the Middle East LOC

Africa LOC

China GPE

Mongolia GPE

South America LOC

Atlantic LOC

Indian NORP

Pacific Oceans NORP

March-May 2018 DATE

March-May DATE

2.34 CARDINAL

the 20th century DATE

46.4 CARDINAL

F PRODUCT

fifth ORDINAL

March-May DATE

139-year DATE

Five CARDINAL

March-May 2018 DATE

nine CARDINAL

continental ORG

1910 DATE

Europe LOC

South America LOC

third ORDINAL

March-May DATE

March-May DATE

1.15 CARDINAL

the 20th century DATE

61.0 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

Year-to-Date DATE

January-May 2018 DATE

1.39 CARDINAL

the 20th century DATE

55.5 CARDINAL

F PRODUCT

2010 DATE

fourth ORDINAL

January-May DATE

139-year DATE

2018 year DATE

0.59 CARDINAL

2016 DATE

2.18 CARDINAL

the 20th century DATE

42.8 CARDINAL

F PRODUCT

2002 DATE

fifth ORDINAL

January-May DATE

1.08 CARDINAL

the 20th century DATE

60.8 CARDINAL

fifth ORDINAL

January-May DATE

1880-2018 DATE

Nigeria GPE

Treasury ORG

EFCC ORG

Theodore Orji PERSON

N521 billion MONEY

Britain GPE

42 MONEY

Penn State ORG

Craig Meyers PERSON

NewsNation ORG

WTAJ ORG

the Journal of Medical Virology ORG

the Penn State College of Medicine ORG

Listerene PERSON

at least 30 seconds TIME

11-year-olds ORDINAL

Oregon GPE

Hundreds CARDINAL

Republican NORP

GOP Voters ORG

Jill Biden PERSON

Shell Corporation That Legally Owns Epstein ORG

the Clinton Foundation ORG

Mike Pence PERSON

Philadelphia GPE

Europe LOC

Spain GPE

France GPE

Germany GPE

Belgium GPE

the Netherlands GPE

Sweden GPE

Sunday DATE

Sweden GPE

100,000 CARDINAL

56.40 CARDINAL

U.K. GPE

69.60 CARDINAL

Spain GPE

60.88 CARDINAL

Italy GPE

58.16 CARDINAL

Sunday DATE

Johns Hopkins University ORG

U.K. GPE

fourth ORDINAL

Spain GPE

Italy GPE

sixth ORDINAL

eighth ORDINAL

Europe LOC

two CARDINAL

Russia GPE

Trump ORG

2020 DATE

Candace Owens PERSON

Nicola Sturgeon's ORG

Britain GPE

today DATE

Trump ORG

Cheyenne GPE

Eugenie PERSON

ninth ORDINAL

the U.S. Centers for Disease Control and Prevention ORG

CDC ORG

the Vaccine Adverse Event Reporting System ORG

more than 10 years DATE

at least 108 CARDINAL

today DATE

today DATE

first ORDINAL

U.S. GPE

2003 DATE

the Associate Press ORG

Anne Schuchat PERSON

CDC ORG

National Center for Immunization and Respiratory Diseases ORG

VAERS ORG

at least 108 CARDINAL

2003 DATE

96 CARDINAL

MMR ORG

first ORDINAL

1963 DATE

U.S. GPE

the years DATE

first ORDINAL

1963 DATE

the United States GPE

approximately 98% PERCENT

the International Medical Council on Vaccination ORG

first ORDINAL

**IMCV PERSON** 

Christian NORP

One Year DATE

Wisconsin Election Commission ORG

Ballots PERSON

BRAZIL GPE

AstraZeneca ORG

Oxford University ORG

Anvisa ORG

one CARDINAL

three CARDINAL

Brazil GPE

Oxford University ORG

the UK Government ORG

100 million CARDINAL

US GPE

September 6 DATE

India GPE

South Africa GPE

AstraZeneca ORG

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Europe LOC
Inclusive Vaccine Alliance ORG
400 million CARDINAL
European NORP
UK GPE
30 million CARDINAL
BioNtech/Pfizer ORG
60 million CARDINAL
Valneva - in PERSON
six CARDINAL
UK GPE
Patrick Vallance PERSON
next spring DATE
the Commons and Lords National Security Strategy Committee ORG
AstraZeneca ORG
the turn of the year DATE
UK GPE
Vaccine Taskforce FAC
Kate Bingham PERSON
UK GPE
four CARDINAL
six CARDINAL
three CARDINAL
one CARDINAL
two CARDINAL
Covid PERSON
26,688 CARDINAL
UK GPE
789,229 CARDINAL
191 CARDINAL
today DATE
May 2018 DATE
fourth ORDINAL
the month of May DATE
NOAA ORG
1880 DATE
March-May DATE
fourth ORDINAL
monthly DATE
NOAA ORG
National Centers for Environmental Information ORG
NOAA ORG
May 2018 DATE
May DATE
1.44 CARDINAL
F PRODUCT
the 20th century DATE
58.6 CARDINAL
F PRODUCT
```

fourth ORDINAL

May DATE

1880-2018 DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.58 CARDINAL

May 2018 DATE

the 42nd consecutive May DATE

the 401st consecutive month DATE

the 20th century DATE

North America LOC

Europe LOC

Asia LOC

Atlantic LOC

Pacific LOC

Indian Oceans NORP

May DATE

Canada GPE

Atlantic Ocean LOC

Greenland GPE

May DATE

2.05 CARDINAL

the 20th century DATE

52.0 CARDINAL

F PRODUCT

May DATE

2011 DATE

2013 DATE

seventh ORDINAL

May DATE

139-year DATE

May 2018 DATE

U.S. GPE

Europe LOC

5.4 CARDINAL

Canada GPE

Russia GPE

5.4 CARDINAL

Europe LOC

May DATE

continental ORG

1910 DATE

4.97 CARDINAL

2003 DATE

1.66 CARDINAL

F PRODUCT

May 2018 DATE

first ORDINAL

May DATE

3.6 CARDINAL

F PRODUCT

2.0 CARDINAL

European NORP

May DATE

Asia LOC

May DATE

2009 DATE

May DATE

1.19 CARDINAL

the 20th century DATE

monthly DATE

61.3 CARDINAL

fourth ORDINAL

May DATE

years 2014-2018 DATE

five CARDINAL

2016 CARDINAL

May DATE

1.39 CARDINAL

May 2018 DATE

May DATE

Arctic LOC

second ORDINAL

40-year DATE

420,000 square miles QUANTITY

8.1 percent PERCENT

1981-2010 DATE

the National Snow and Ice Data Center(link ORG

NOAA ORG

NASA ORG

May Arctic LOC

2016 DATE

Arctic LOC

Bering GPE

Chukchi Seas PERSON

May DATE

330,000 square miles QUANTITY

8.6 percent PERCENT

1981-2010 DATE

third ORDINAL

May DATE

Antarctic sea LOC

May DATE

the Weddell Sea LOC

Seasonal DATE

March-May 2018 DATE

March-May DATE

1.48 CARDINAL

the 20th century DATE

56.7 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

1880-2018 DATE

0.45 CARDINAL

F cooler PRODUCT

2016 DATE

March-May DATE

2014 DATE

three-month DATE

Europe LOC

the Middle East LOC

Africa LOC

China GPE

Mongolia GPE

South America LOC

Atlantic LOC

Indian NORP

Pacific Oceans NORP

March-May 2018 DATE

March-May DATE

2.34 CARDINAL

the 20th century DATE

46.4 CARDINAL

F PRODUCT

fifth ORDINAL

March-May DATE

139-year DATE

Five CARDINAL

March-May 2018 DATE

nine CARDINAL

continental ORG

1910 DATE

Europe LOC

South America LOC

third ORDINAL

March-May DATE

March-May DATE

1.15 CARDINAL

the 20th century DATE

61.0 CARDINAL

F PRODUCT

fourth ORDINAL

March-May DATE

Year-to-Date DATE

January-May 2018 DATE

1.39 CARDINAL

the 20th century DATE

55.5 CARDINAL

```
F PRODUCT
```

2010 DATE

fourth ORDINAL

January-May DATE

139-year DATE

2018 year DATE

0.59 CARDINAL

2016 DATE

2.18 CARDINAL

the 20th century DATE

42.8 CARDINAL

F PRODUCT

2002 DATE

fifth ORDINAL

January-May DATE

1.08 CARDINAL

the 20th century DATE

60.8 CARDINAL

fifth ORDINAL

January-May DATE

1880-2018 DATE

This Labor Day DATE

Tory PERSON

Pope PERSON

80 CARDINAL

Child Trafficking, Fraud — Conservative WORK\_OF\_ART

BRAZIL GPE

AstraZeneca ORG

Oxford University ORG

Anvisa ORG

one CARDINAL

three CARDINAL

Brazil GPE

Oxford University ORG

the UK Government ORG

100 million CARDINAL

US GPE

September 6 DATE

India GPE

South Africa GPE

AstraZeneca ORG

Europe LOC

Inclusive Vaccine Alliance ORG

400 million CARDINAL

European NORP

UK GPE

30 million CARDINAL

BioNtech/Pfizer ORG

60 million CARDINAL

Valneva - in PERSON

six CARDINAL

UK GPE

Patrick Vallance PERSON

next spring DATE

the Commons and Lords National Security Strategy Committee ORG

AstraZeneca ORG

the turn of the year DATE

UK GPE

Vaccine Taskforce FAC

Kate Bingham PERSON

UK GPE

four CARDINAL

six CARDINAL

three CARDINAL

one CARDINAL

two CARDINAL

Covid PERSON

26,688 CARDINAL

UK GPE

789,229 CARDINAL

191 CARDINAL

today DATE

Vigil PERSON

13-year-old DATE

Nicola Sturgeon's ORG

The Novel Coronavirus Infecting ORG

China GPE

Army ORG

Yahaya Bello PERSON

Massive International Study Shows Countries ORG

79% PERCENT

Lower Mortality Rate ORG

100,000 CARDINAL

American NORP

2030 DATE

Barack Obama PERSON

Capitol Shooting Exposed ORG

Hollywood GPE

Canadian NORP

Programmable Hydraulic Guillotines WORK OF ART

Canada GPE

COVID-19 ORG

Newt Gingrich PERSON

Arctic LOC

Winter Sea Ice Drops LOC

Britain GPE

Windrush PERSON

SNUBBED ORG

week DATE

Commonwealth ORG

Biden Plans PERSON

Move Social Security Eligibility ORG

75 CARDINAL

today DATE

James I about PERSON

four centuries ago DATE

Duane Carr PERSON

the University of Tennessee College of Medicine ORG

Victor Buhler PERSON

St. Joseph Hospital FAC

Kansas City GPE

thousands CARDINAL

Sheldon Sommers PERSON

Pathologist NORP

Lenox Hill Hospital ORG

New York GPE

Brigitte ORG

Youtube PERSON

German NORP

Rich White's PERSON

Smoke Screens PERSON

Jan Zeldenrust PERSON

Dutch NORP

the Government of Holland ORG

1951 - 1984 DATE

1980 DATE

Dutch NORP

these days DATE

Coal ORG

First ORDINAL

emphysema PERSON

4 CARDINAL

5 CARDINAL

4 CARDINAL

8) CARDINAL

first ORDINAL

emphysema PERSON

secondly ORDINAL

emphysema PERSON

Delhi GPE

'Delhi GPE

Delhi GPE

Indian NORP

Delhi GPE

Indian NORP

Delhi GPE

'Delhi GPE

emphysema PERSON

one CARDINAL

BioFlex ORG

firstly ORDINAL

secondly ORDINAL

the University of Arkansas ORG

Villanova University ORG

Marquette University ORG

more than 500 CARDINAL

U.S. GPE

Canadian NORP

Scot Burton PERSON

the Sam M. Walton College of Business ORG

two CARDINAL

Alex Trebek Becomes PERSON

U.S. GPE

24% PERCENT

597 CARDINAL

a single year DATE

China GPE

the World Health Organization ORG

the U.S. National Institutes of Health ORG

more than 2 million CARDINAL

Wuhan GPE

China GPE

the United States GPE

COVID-19 ORG

Fauci PERSON

Coronavirus PERSON

Art Moore PERSON

February 1, 2021 DATE

Tweet Gab Share PERSON

Anthony Fauci PERSON

National Institute of Allergy and Infectious Diseases ORG

Thursday, July 30, 2020 DATE

Washington GPE

D.C. GPE

White House ORG

Tia Dufour PERSON

Feb. 8, 2021: DATE

USA Today ORG

March DATE

April 2020 DATE

Chinese NORP

Fadela Chaib PERSON

The Scripps Research Institute ORG

FactCheck.org ORG

Feb. 7, 2020 DATE

the last 20 years DATE

Timothy Sheahan PERSON

the University of North Carolina ORG

Chapel Hill FAC

96% PERCENT

China GPE

the World Health Organization ORG

the U.S. National Institutes of Health ORG

more than 2 million CARDINAL

Wuhan GPE

China GPE

the United States GPE

Newsweek ORG

April 2020 DATE

NIH ORG

Anthony Fauci PERSON

more than 200 CARDINAL

Obama PERSON

2014 DATE

U.S. GPE

Fauci ORG

2017 DATE

April 2020 DATE

the Wuhan Institute of Virology ORG

China GPE

Fauci's National Institute of Allergy and Infectious Diseases ORG

Steve Hilton PERSON

Fox News ORG

The Next Revolution WORK OF ART

a decade ago DATE

Yunnan province GPE

Wuhan GPE

1,000 miles QUANTITY

Hilton PERSON

Sunday DATE

Wuhan GPE

February 2020 DATE

Yunnan GPE

COVID-19 PERSON

Hilton PERSON

NIH ORG

GenBank ORG

only one CARDINAL

COVID-19 ORG

a decade ago DATE

Yunnan GPE

Hilton GPE

Wuhan GPE

Yunnan GPE

GenBank ORG

Yunnan GPE

Wuhan GPE

Hilton PERSON

two CARDINAL

two CARDINAL

COVID-19 PERSON

NAID ORG

a Nov. 30, 2017 DATE

NAID ORG

Hilton PERSON

a decade ago DATE

NAID ORG

Wuhan GPE

first ORDINAL

COVID-19 ORG

the fall of 2019 DATE

USCIS ORG

Trump ORG

24% PERCENT

597 CARDINAL

a single year DATE

Biden PERSON

Russia GPE

SolarWinds ORG

Ohio GPE

163 CARDINAL

George Soros For PERSON

The World Health Organization ORG

months DATE

WHO ORG

WHO ORG

34 CARDINAL

Monday October 5th DATE

Michael Ryan PERSON

WHO ORG

roughly 10% PERCENT

around 35 million CARDINAL

Margaret Harris PERSON

WHO ORG

WHO ORG

Ryan PERSON

roughly 7.8 billion CARDINAL

10% PERCENT

780 million CARDINAL

1,061,539 CARDINAL

0.14% PERCENT

0.14% PERCENT

24 CARDINAL

WHO ORG

3.4% PERCENT

March DATE

Covid PERSON

IFR ORG

0.14% PERCENT

Covid PERSON

Dr Ryan's PERSON

WHO ORG

Covid NORP

US GPE

CDC ORG

The New York Times ORG

Christian NORP

One Year DATE

The Prices Of Food Items Are Coming Down ORG

Garba Shehu PERSON

1bn MONEY

next two years DATE

Europe LOC

Rick Scott PERSON

Democrats NORP

Liberal States GPE

Order Deeming Church An Essential Business ORG

21 Million CARDINAL

China GPE

Coronavirus Outbreak GPE

Alaska GPE

Permafrost Is Thawing WORK OF ART

Berger PERSON

NC GPE

12 CARDINAL

Graphs Show Mask Mandates Do Nothing To Stop COVID EVENT

One CARDINAL

five CARDINAL

USCIS ORG

Vladimir Putin PERSON

Europe LOC

Vitamin ORG

Justine Greening PERSON

Hayes PERSON

GEORGIA GPE

Tesco ORG

119 CARDINAL

last year DATE

New Yorkers NORP

Breonna Taylor PERSON

SHE ORG

first ORDINAL

COVID-19 ORG

COVID-19 ORG

RNA ORG

COVID-19 ORG

Moderna GPE

RNA ORG

thirty years ago DATE

5 CARDINAL

```
two CARDINAL
RNA ORG
Coronavirus LOC
U.S. GPE
Chicago GPE
hundreds of millions MONEY
Americans NORP
the end of the year DATE
Bill Gates PERSON
WHO ORG
years DATE
Gates ORG
7 billion CARDINAL
RNA ORG
Gates ORG
DNA/RNA ORG
Vaccines NORP
two CARDINAL
COVID-19 ORG
a ton OUANTITY
two CARDINAL
RNA ORG
one CARDINAL
RNA ORG
CEPI ORG
RNA ORG
nearly a decade DATE
one CARDINAL
COVID ORG
first ORDINAL
RNA ORG
Moderna GPE
RNA ORG
RNA ORG
RNA ORG
RNA ORG
COVID ORG
first ORDINAL
RNA ORG
first ORDINAL
Bill Gates PERSON
Newt Gingrich PERSON
```

Cette fonction a pour but d'ajouter le type d'entité de chaque token détecté à coté de chaque entité détéctée dans le texte qu'elle prend en entrée, et retourne le texte modifié avec les informations ajoutées sur les entités nommées.

```
def add_entity_name(text):
    if text is None or pd.isna(text):
        return ""
```

```
doc = nlp(text)
    # Créer une liste pour stocker les nouveaux tokens
    nouveaux tokens = []
    # Parcourir les tokens et ajouter des informations selon le type
d'entité
    for token in doc:
        if token.ent type :
            # Ajouter le nom de l'entité (type d'entité) à côté de
l'entité
            nouveaux_tokens.append(f"{token.text}
({token.ent type })")
        else:
            nouveaux tokens.append(token.text)
    # Reconstruire le texte avec les informations ajoutées
    nouveau texte = " ".join(nouveaux tokens)
    return nouveau texte
Application de la fonction précédente à la colonne "title" avec la méthode apply
# Assurez-vous que la colonne "title" contient des chaînes de
caractères
dftrain['title'] = dftrain['title'].astype(str)
# Appliquer la fonction à la colonne "title"
dftrain['title'] = dftrain['title'].apply(add entity name)
print(dftrain['title'])
615
        JK (ORG) Rowling (ORG) Confirms (ORG) Stance (...
        MI (PERSON) Sec (PERSON) of State (ORG) Offici...
1303
1232
        What science can tell us about the links betwe...
2022
        Sarah (PERSON) Parker (PERSON) leaves legacy o...
        Current Actual Election Result Update: Presid...
287
1006
        Re : Meeting the need for isolation space for ...
1543
        Breaking: Breonna (PERSON) Taylor (PERSON) 's...
853
        A quote from Politifact : Gates never said tha...
296
        Before This Election , Newt (PERSON) Gingrich ...
1325
         COVID19 PCR Tests are Scientifically Meaningless
Name: title, Length: 844, dtype: object
On scinde les données de la colonne "title" en jeu d'apprentissage et jeu de test
X=dftrain['title']
X train, X test, y train, y test=train test split(X, y, test size =
0.2, random state=8)
print("X train is \n", X_train.shape)
print("y train is \n",y train.shape)
```

```
print("X_test is \n", X_test.shape)
print("y_test is \n", y_test.shape)

X_train is
  (675,)
y_train is
  (675,)
X_test is
  (169,)
y_test is
  (169,)
```

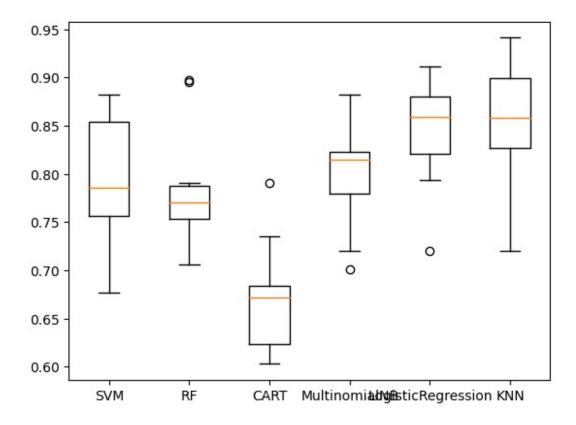
Ici, c'est une étape importante, on va tester différents classifieurs, pour chacun des classifieurs, on va appliquer le prétraitement + Vectorisation TfIdf, et on applique une cross\_val\_score avec un Kfold de 10 fois, par la suite on stocke dans une liste all\_results la moyenne des accuracy + l'écart type et on la trie par ordre décroissant de moyenne d'accuracy et d'écart type. on remarque que les 2 meilleurs sont SVM et RF qu'on va séléctionner pour leur appliquer le GridSearch sur les paramètres des prétraitements + leurs hyperparamètres pour pouvoir choisir le meilleur.

```
np.random.seed(42) # Set the random seed for NumPv
score = 'accuracy'
seed = 7
allresults = []
results = []
names = []
# Liste des modèles à tester
models = [
    ('MultinomialNB', MultinomialNB()),
    ('LogisticRegression', LogisticRegression(random state=42))
1
models.append(('KNN', KNeighborsClassifier()))
models.append(('CART', DecisionTreeClassifier(random state=42)))
models.append(('RF', RandomForestClassifier(random state=42)))
models.append(('SVM', SVC(random_state=42)))
# Création d'un pipeline pour chaque modèle
pipelines = []
for name, model in models:
    pipeline = Pipeline([
        ('normalize', TextNormalizer()),
        ('tfidf', TfidfVectorizer()),
        (name, model)
```

```
1)
    pipelines.append((name,pipeline))
    #pipeline.fit(X_train,y_train)
all results=[]
scores=[]
for p in pipelines:
    print(p[1])
    # cross validation en 10 fois
    kfold = KFold(n splits=10, random state=seed, shuffle=True)
    print ("Evaluation de ",p)
    start time = time.time()
    # application de la classification
    cv results = cross val score(p[1],X train,y train, cv=kfold,
scoring=score)
    #print("Pour le classifieur",p[0],"on a un score
de",cv_results.mean(),"et un écart type de",cv_results.std())
    scores.append(cv_results)
    all results.append((p[0],cv results.mean(),cv results.std()))
    end time = time.time()
print("all resultats", all results)
all results = sorted(all results, key=lambda x: (-x[1], -x[2]))
print("all resultats", all_results)
    # affichage des résultats
#print ('\nLe meilleur resultat : ',max(results))
X train (675,)
y train (675,)
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('MultinomialNB', MultinomialNB())])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('LogisticRegression',
LogisticRegression(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('KNN', KNeighborsClassifier())])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('CART', DecisionTreeClassifier(random_state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                ('RF', RandomForestClassifier(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
```

```
('SVM', SVC(random state=42))])
all resultats [('MultinomialNB', 0.7926031606672521,
0.06432746085775896), ('LogisticRegression', 0.7867427568042142,
0.058944632458151805), ('KNN', 0.6712686567164179,
0.054767770623612606), ('CART', 0.7983977172958736,
0.05136480571971499), ('RF', 0.8444907813871817,
0.054290524553410345), ('SVM', 0.8534460052677787,
0.06256935356011088)]
all resultats [('SVM', 0.8534460052677787, 0.06256935356011088),
('RF', 0.8444907813871817, 0.054290524553410345), ('CART',
0.7983977172958736, 0.05136480571971499), ('MultinomialNB',
0.7926031606672521, 0.06432746085775896), ('LogisticRegression',
0.7867427568042142, 0.058944632458151805), ('KNN', 0.6712686567164179,
0.054767770623612606)1
On affiche les boites à moustache pour mieux visualiser les résultats
import matplotlib.pyplot as plt
fig = plt.figure()
fig.suptitle('Comparaison des algorithmes')
ax = fig.add subplot(111)
plt.boxplot(scores)
ax.set xticklabels(name[0] for name in all results)
[Text(1, 0, 'SVM'),
 Text(2, 0, 'RF'),
 Text(3, 0, 'CART'),
 Text(4, 0, 'MultinomialNB'),
 Text(5, 0, 'LogisticRegression'),
 Text(6, 0, 'KNN')]
```

## Comparaison des algorithmes



On a un pipeline pour chaque prétraitement différent, on essaye pas mal (miniscule, lemmatisation, miniscule + lemmatisation..) et on stocke le fit\_transorm de nos X\_train, X\_test sur les pipelines dans des listes qui vont contenir tous les fit\_transform des pipelines pour chaque classifieur, par la suite on parcourt ces listes là, on itère dessus, et chaque élement de la liste (train) va passer par le GridSearch et puis on predict sur son corresapondant dans liste (test).

```
getstemmer=False, removedigit=False)),
                     ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowStopstem = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=True, removedigit=False)),
                     ('count vectorizer',
CountVectorizer(lowercase=False))])
# pipeline de l'utilisation de TfidfVectorizer avec differents pre-
traitements
TFIDF brut = Pipeline ([('cleaner', TextNormalizer()),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowcase = Pipeline([('cleaner',
TextNormalizer(removestopwords=False,lowercase=True,
getstemmer=False, removedigit=False)),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowStop = Pipeline([('cleaner'
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=False,removedigit=False)),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowStopstem = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=True,removedigit=False)),
                     ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
# Liste de tous les modeles à tester
all models = [
    ("CV brut", CV brut),
    ("CV lowcase", CV lowcase),
    ("CV lowStop", CV lowStop),
    ("CV_lowStopstem", CV_lowStopstem),
    ("TFIDF_lowcase", TFIDF_lowcase),
("TFIDF_lowStop", TFIDF_lowStop),
    ("TFIDF_lowStopstem", TFIDF_lowStopstem),
    ("TFIDF_brut", TFIDF_brut)
1
```

```
X train title SVC = []
X test title SVC = []
X train title RandomForestClassifier = []
X test title RandomForestClassifier = []
for name, pipeline in all models :
X train title SVC.append(pipeline.fit transform(X train).toarray())
    X_test_title_SVC.append(pipeline.transform(X_test).toarray())
X train title RandomForestClassifier.append(pipeline.fit transform(X t
rain).toarray())
X test title RandomForestClassifier.append(pipeline.transform(X test).
toarray())
models = {
    'SVC': SVC(random state=42),
    'RandomForestClassifier': RandomForestClassifier(random state=42)
}
params = \{'SVC': [\{'C': [0.001, 0.01, 0.1, 1,2,5,7,10]\},
             {'gamma': [0.001, 0.01, 0.1,0.2,0.3,0.5,0.7,1]},
             {'kernel': ['linear', 'rbf']}],
    'RandomForestClassifier': [{'n estimators': [10, 50, 100, 200,
3001},
                              {'max features': ['auto', 'sqrt',
'log2']}]
for model name, model in models.items():
    score='accuracy'
    X train title = eval('X train title ' + model name)
    X_test_title = eval('X_test_title ' + model name)
    for i in range (len(X train title)):
      grid search = GridSearchCV(model, params[model name], n jobs=-1,
verbose=1,scoring=score)
      print("grid search fait")
      print("X_train",X_train_title[i].shape)
      print("y_train",y_train.shape)
      grid_search.fit(X_train_title[i],y_train)
      print ('meilleur score %0.3f'%(grid search.best score ),'\n')
```

```
print ('meilleur estimateur',grid_search.best_estimator_,'\n')
      y pred = grid search.predict(X test title[i])
     MyshowAllScores(y_test,y_pred)
      print("Ensemble des meilleurs paramètres :")
      best parameters = grid search.best estimator .get params()
      for param dict in params[model name]:
        for param name, param value in param dict.items():
           print("\t%s: %r" % (param name,
best parameters[param name]))
y train (675,)
y test (169,)
X test (169,)
grid search fait
X train (675, 5737)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.828
meilleur estimateur SVC(gamma=0.5, random state=42)
Accuracy: 0.911
Classification Report
              precision
                          recall f1-score
                                             support
                         0.98795
                                                  83
       FALSE
               0.85417
                                   0.91620
                         0.83721
        TRUE
               0.98630
                                   0.90566
                                                  86
                                   0.91124
                                                 169
   accuracy
   macro avg
               0.92023
                         0.91258
                                   0.91093
                                                  169
weighted avg
               0.92141
                         0.91124
                                   0.91084
                                                 169
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 0.5
     kernel: 'rbf'
grid search fait
X train (675, 4858)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.831
meilleur estimateur SVC(gamma=0.5, random state=42)
Accuracy: 0.911
Classification Report
              precision
                          recall f1-score
                                             support
       FALSE
               0.85417 0.98795
                                   0.91620
                                                   83
```

```
TRUE
                0.98630
                          0.83721
                                    0.90566
                                                   86
                                    0.91124
                                                  169
    accuracy
                0.92023
                          0.91258
                                    0.91093
                                                  169
   macro avq
weighted avg
                0.92141
                          0.91124
                                    0.91084
                                                  169
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 0.5
     kernel: 'rbf'
grid search fait
X train (675, 4740)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.828
meilleur estimateur SVC(gamma=0.5, random state=42)
Accuracy: 0.935
Classification Report
              precision
                           recall f1-score
                                              support
                0.90909
                          0.96386
                                    0.93567
       FALSE
                                                   83
        TRUE
                0.96296
                          0.90698
                                    0.93413
                                                   86
    accuracy
                                    0.93491
                                                  169
                0.93603
                          0.93542
                                    0.93490
                                                  169
   macro avq
weighted avg
                0.93651
                          0.93491
                                    0.93489
                                                  169
Ensemble des meilleurs paramètres :
     C: 1.0
     gamma: 0.5
     kernel: 'rbf'
grid search fait
X train (675, 3619)
v train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.828
meilleur estimateur SVC(gamma=0.5, random_state=42)
Accuracy: 0.923
Classification Report
              precision
                           recall f1-score
                                              support
                0.88889
                          0.96386
                                    0.92486
       FALSE
                                                   83
                0.96203
        TRUE
                          0.88372
                                    0.92121
                                                   86
                                    0.92308
                                                  169
    accuracy
```

```
macro avg 0.92546 0.92379 0.92303 169
weighted avg 0.92611 0.92308 0.92300 169

Ensemble des meilleurs paramètres :
        C: 1.0
        gamma: 0.5
        kernel: 'rbf'
grid search fait
```

X\_train (675, 4858) y\_train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.831

meilleur estimateur SVC(C=2, random\_state=42)

Accuracy: 0.893

Classification Report

support	f1-score	recall	precision	
83 86	0.89286 0.89412	0.90361 0.88372	0.88235 0.90476	FALSE TRUE
169 169 169	0.89349 0.89349 0.89350	0.89367 0.89349	0.89356 0.89376	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

C: 2

gamma: 'scale' kernel: 'rbf' grid search fait X\_train (675, 4740) y\_train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.822

meilleur estimateur SVC(C=1, random\_state=42)

Accuracy : 0.858

Classification Report

support	f1-score	recall	precision	
83 86	0.85366 0.86207	0.84337 0.87209	0.86420 0.85227	FALSE TRUE
169 169 169	0.85799 0.85786 0.85794	0.85773 0.85799	0.85824 0.85813	accuracy macro avg weighted avg

```
Ensemble des meilleurs paramètres :
     C: 1
     gamma: 'scale'
     kernel: 'rbf'
grid search fait
X_train (675, 3619)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.813
meilleur estimateur SVC(C=2, random state=42)
Accuracy: 0.876
Classification Report
                           recall f1-score
              precision
                                              support
       FALSE
                0.87805
                          0.86747
                                    0.87273
                                                   83
        TRUE
                0.87356
                          0.88372
                                    0.87861
                                                   86
                                    0.87574
                                                  169
    accuracy
                0.87581
                          0.87560
                                    0.87567
                                                  169
   macro avg
weighted avg
               0.87577
                          0.87574
                                    0.87572
                                                  169
Ensemble des meilleurs paramètres :
     C: 2
     gamma: 'scale'
     kernel: 'rbf'
grid search fait
X_train (675, 5737)
y train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.831
meilleur estimateur SVC(C=2, random state=42)
Accuracy: 0.876
Classification Report
              precision
                           recall f1-score
                                              support
       FALSE
                0.87805
                          0.86747
                                    0.87273
                                                   83
        TRUE
                0.87356
                          0.88372
                                    0.87861
                                                   86
                                    0.87574
                                                  169
    accuracy
                0.87581
                          0.87560
                                    0.87567
                                                  169
   macro avg
weighted avg
                0.87577
                          0.87574
                                    0.87572
                                                  169
Ensemble des meilleurs paramètres :
```

C: 2

gamma: 'scale'

kernel: 'rbf'

grid search fait

X\_train (675, 5737)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.813

meilleur estimateur RandomForestClassifier(n\_estimators=200, random\_state=42)

Accuracy: 0.905

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.90361 0.90698	0.90361 0.90698	0.90361 0.90698	83 86
accuracy macro avg weighted avg	0.90530 0.90533	0.90530 0.90533	0.90533 0.90530 0.90533	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 200

max\_features: 'sqrt'

grid search fait

X\_train (675, 4858)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.803

meilleur estimateur RandomForestClassifier(n\_estimators=10, random\_state=42)

Accuracy: 0.888

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.90000 0.87640	0.86747 0.90698	0.88344 0.89143	83 86
-	0107010	0.30030		
accuracy			0.88757	169
macro avg	0.88820	0.88722	0.88743	169
weighted avg	0.88799	0.88757	0.88750	169

Ensemble des meilleurs paramètres :

n\_estimators: 10

max\_features: 'sqrt'

grid search fait

X\_train (675, 4740)

y\_train (675,)
Fitting 5 folds for each of 8 candidates, totalling 40 fits
meilleur score 0.812

meilleur estimateur RandomForestClassifier(n\_estimators=50,
random state=42)

Accuracy: 0.852

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.86250 0.84270	0.83133 0.87209	0.84663 0.85714	83 86
accuracy macro avg weighted avg	0.85260 0.85242	0.85171 0.85207	0.85207 0.85188 0.85198	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 50
max\_features: 'sqrt'

grid search fait

X\_train (675, 3619)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.801

meilleur estimateur RandomForestClassifier(n\_estimators=50, random\_state=42)

Accuracy: 0.846

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.87013 0.82609	0.80723 0.88372	0.83750 0.85393	83 86
accuracy macro avg weighted avg	0.84811 0.84772	0.84547 0.84615	0.84615 0.84572 0.84586	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 50
max features: 'sqrt'

grid search fait

X\_train (675, 4858)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.819

## meilleur estimateur RandomForestClassifier(random\_state=42)

Accuracy: 0.876

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.87805 0.87356	0.86747 0.88372	0.87273 0.87861	83 86
accuracy macro avg weighted avg	0.87581 0.87577	0.87560 0.87574	0.87574 0.87567 0.87572	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 100
max\_features: 'sqrt'

grid search fait X train (675, 4740)

y train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.818

meilleur estimateur RandomForestClassifier(n\_estimators=300, random\_state=42)

Accuracy: 0.864

Classification Report

support	f1-score	recall	precision	
83 86	0.86228 0.86550	0.86747 0.86047	0.85714 0.87059	FALSE TRUE
169 169 169	0.86391 0.86389 0.86391	0.86397 0.86391	0.86387 0.86398	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

n\_estimators: 300
max\_features: 'sqrt'

grid search fait

X train (675, 3619)

y train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.813

meilleur estimateur RandomForestClassifier(n\_estimators=300, random state=42) Accuracy: 0.852

Classification Report

support	f1-score	recall	precision	
83 86	0.84848 0.85549	0.84337 0.86047	0.85366 0.85057	FALSE TRUE
169 169 169	0.85207 0.85199 0.85205	0.85192 0.85207	0.85212 0.85209	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

n\_estimators: 300
max\_features: 'sqrt'

grid search fait X\_train (675, 5737)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.812

meilleur estimateur RandomForestClassifier(n\_estimators=300, random\_state=42)

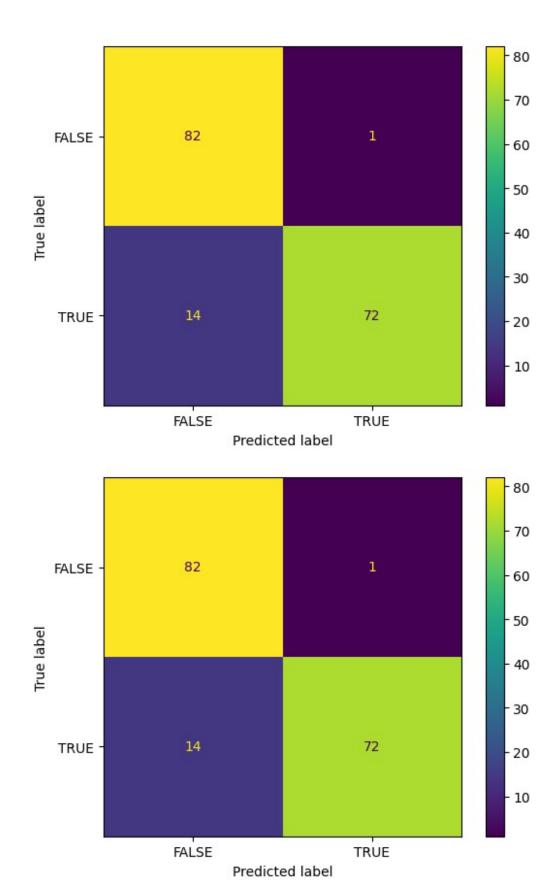
Accuracy: 0.899

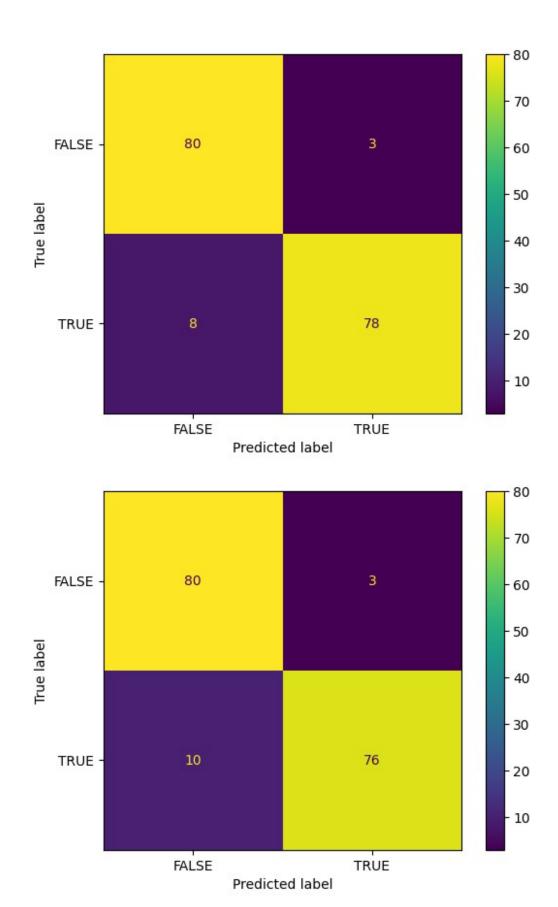
Classification Report

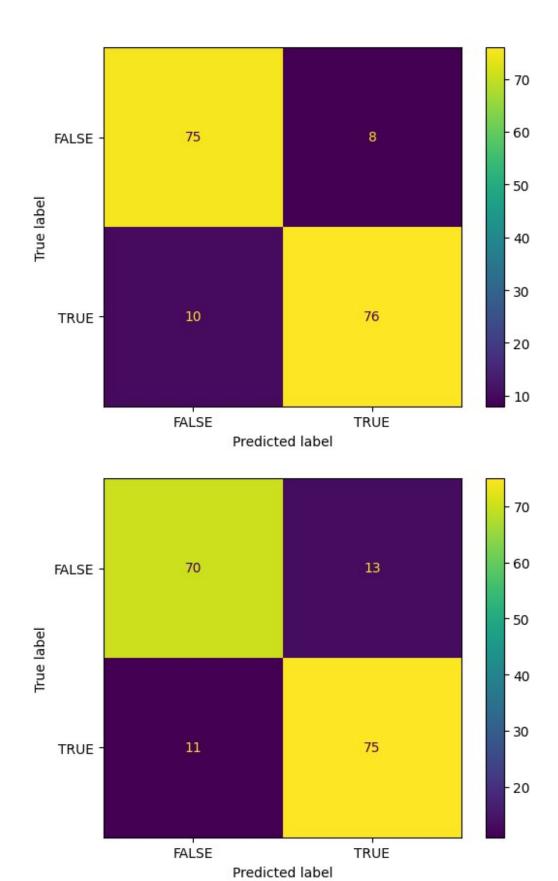
	precision	recall	f1-score	support
FALSE TRUE	0.90244 0.89655	0.89157 0.90698	0.89697 0.90173	83 86
accuracy macro avg weighted avg	0.89950 0.89944	0.89927 0.89941	0.89941 0.89935 0.89939	169 169 169

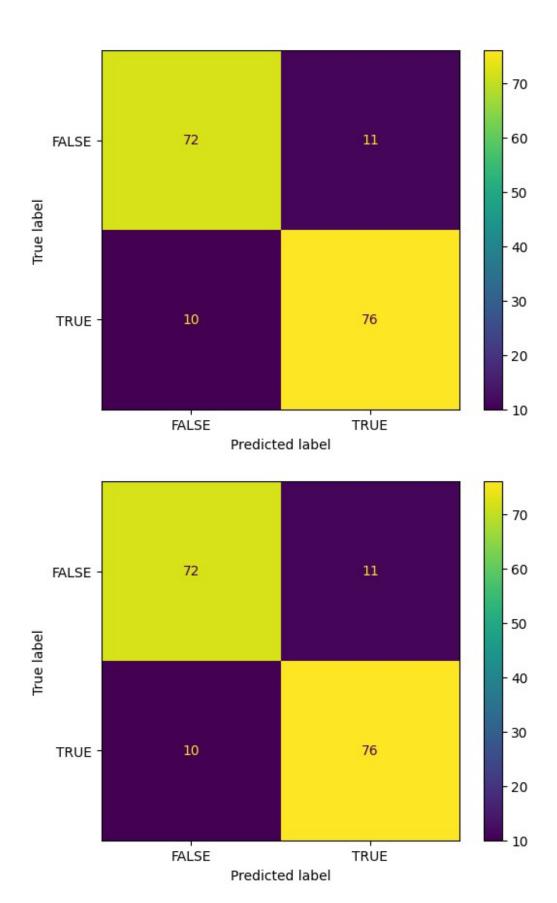
Ensemble des meilleurs paramètres :

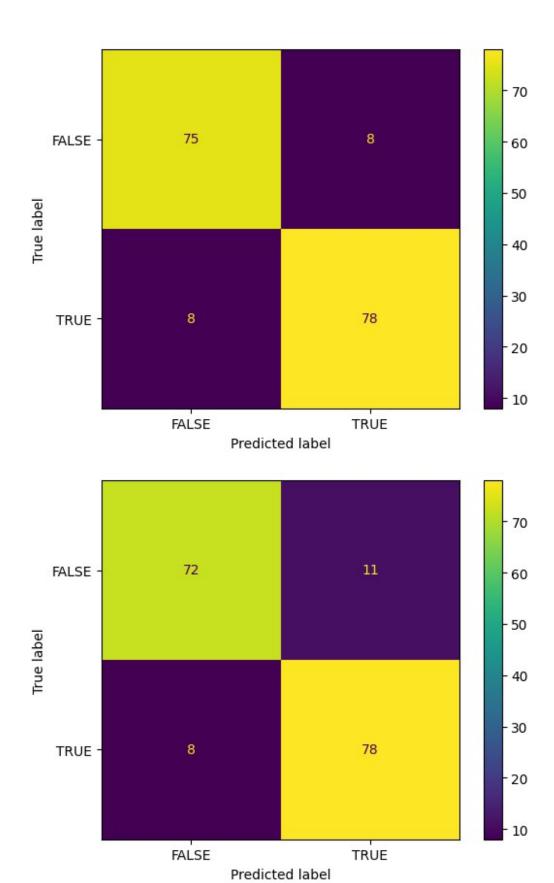
n\_estimators: 300
max\_features: 'sqrt'

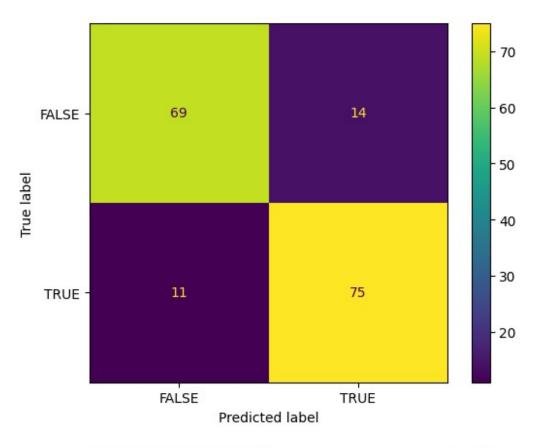


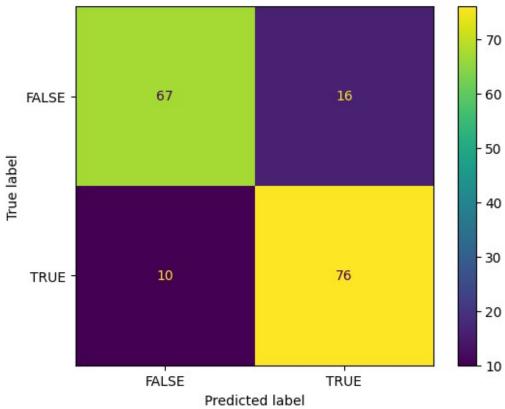


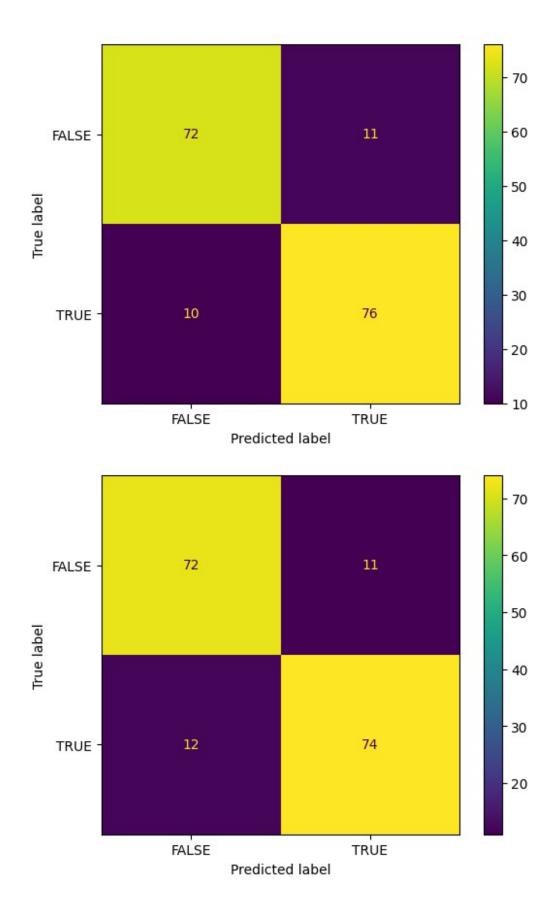


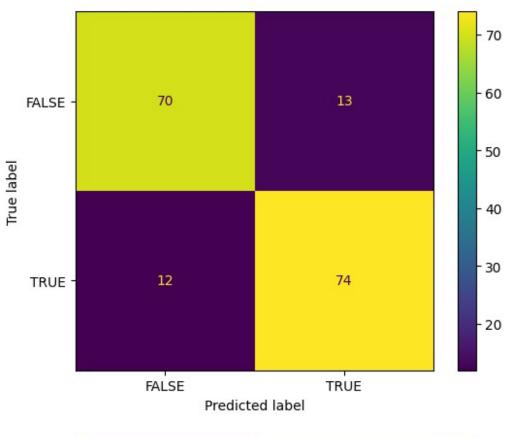


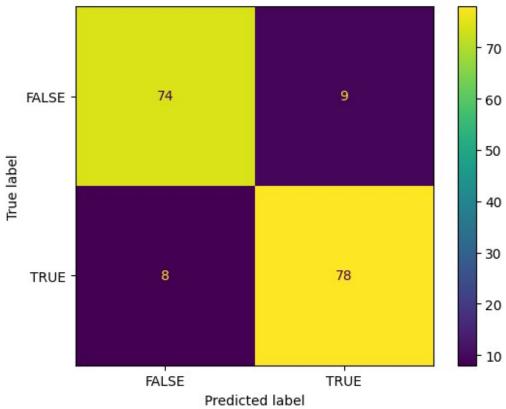












## ##Etape 4 : Classification selon la colonne TEXT\_TITLE + extraction/application d'entités nommées

Tests de Classification pour trouver le meilleur classifieur, les meilleurs paramètres de prétraitement et les meilleurs hyper\_paramètres.

Avant de classifier on applique les entités nommées, Cette fonction utilise Spacy pour extraire les entités nommées à partir d'une liste de textes et les stocke dans une liste pour être affichées avec leurs labels correspondants.

concaténation des cellules "text" et "title"

```
text_title = dftrain.apply(lambda x : '{}
{}'.format(x['text'],x['title']),axis=1)

dftrain['text_title'] = text_title
X_text_title = dftrain.iloc[0:,-1]
X_text_title.reset_index(drop = True, inplace = True)
```

Cette fonction a pour but d'ajouter le type d'entité de chaque token détecté à coté de chaque entité détéctée dans le texte qu'elle prend en entrée, et retourne le texte modifié avec les informations ajoutées sur les entités nommées.

```
texte=dftrain['text title']
#Créer une liste pour stocker les entités nommées
entities = []
# Traiter chaque texte individuellement et ajouter les entités à la
for phrase in texte:
    doc = nlp(phrase)
    for ent in doc.ents:
        entities.append((ent.text, ent.label ))
# Afficher les entités nommées
for entity in entities:
    print(entity[0], entity[1])
Le flux de sortie a été tronqué et ne contient que les 5000 dernières
lignes.
FAC ORG
96 (PERCENT) % PERCENT
China GPE
GPE ORG
U.S. GPE
CARDINAL ORG
CARDINAL ORG
2 CARDINAL
CARDINAL ORG
CARDINAL ORG
Wuhan GPE
```

GPE ORG

China GPE

GPE ORG

GPE ORG

Newsweek ORG

ORG ORG

April DATE

2020 DATE

NIH ORG

ORG ORG

Anthony PERSON

CARDINAL ORG

CARDINAL ORG

200 CARDINAL

CARDINAL ORG

2014 DATE

U.S. GPE

GPE ORG

Fauci PERSON

ORG ORG

2017 DATE

April DATE

2020 DATE

Wuhan GPE

ORG) Institute ORG

China GPE

GPE ORG

Fauci (ORG) 's ORG

ORG ORG

Steve PERSON

Fox ORG

ORG ORG

The (WORK\_OF\_ART) WORK\_OF\_ART

Yunnan GPE

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GPE ORG

Wuhan GPE

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1,000 CARDINAL

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miles PERSON

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Sunday DATE

Wuhan GPE

GPE ORG

February DATE

2020 DATE

Yunnan GPE

GPE ORG

COVID-19 PERSON

Hilton PERSON

NIH ORG

ORG ORG

GenBank ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

COVID-19 ORG

ORG ORG

Yunnan GPE

GPE ORG

Hilton GPE

GPE ORG

Wuhan GPE

GPE ORG

Yunnan GPE

GPE ORG

GenBank ORG

ORG ORG

Yunnan GPE

GPE ORG

Wuhan GPE

GPE ORG

Hilton PERSON

two CARDINAL

CARDINAL ORG

two CARDINAL

CARDINAL ORG

COVID-19 PERSON

NAID ORG

30 CARDINAL

2017 DATE

NAID ORG

ORG ORG

Hilton PERSON

NAID ORG

ORG ORG

Wuhan GPE

GPE ORG

first ORDINAL

COVID-19 ORG

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2019 DATE

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GPE ORG

first ORDINAL

Today DATE

Nicola ORG

ORG ORG

Dandridge PERSON

CBE PERSON

ORG ORG

Chris PERSON

Dame Julia PERSON

first ORDINAL

100 CARDINAL

years DATE

5 CARDINAL

years DATE

David PERSON

Greg PERSON

Clark PERSON

OECD ORG

ORG ORG

Graduate PERSON

2007 DATE

87.5 (PERCENT) % PERCENT

UK GPE

GPE ORG

HE GPE

GPE ORG

4 CARDINAL

CARDINAL ORG

10 CARDINAL

CARDINAL ORG

Unfinished PERSON

months DATE

3 CARDINAL

CARDINAL ORG

firstly ORDINAL

2020 DATE

2009 DATE

first ORDINAL

2012 DATE

HEPI ORG

ORG ORG

BBC ORG

ORG ORG

ORG ORG

ComRes ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

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ORG ORG

March DATE

Department (ORG ORG

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BIS ORG
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ORG) DfE ORG
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Department (ORG ORG
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National (ORG) Student (ORG) Survey (ORG ORG
US GPE
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CARDINAL ORG
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three CARDINAL
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3 CARDINAL
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2006 DATE
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PERCENT ORG
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22,000 MONEY
CBI ORG
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Skills (ORG ORG
47 (PERCENT) % PERCENT
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David PERSON
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English LANGUAGE
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REF ORG
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ORG ORG
2010 DATE
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ORG) Department ORG

Skills PRODUCT

CARDINAL ORG

TEF ORG

ORG ORG

UK GPE

GPE ORG

2015 DATE

ORG ORG

2:1 CARDINAL

QUANTITY ORG

QUANTITY ORG

2:1s CARDINAL

CARDINAL ORG

UK GPE

GPE ORG

300 (PERCENT) % PERCENT

1990s DATE

70 DATE

First ORDINAL

2:1 CARDINAL

CARDINAL ORG

CARDINAL ORG

5 CARDINAL

years DATE

PERCENT ORG

47 DATE

mid-1990s DATE

2013 DATE

14 CARDINAL

CARDINAL ORG

PERCENT ORG

second ORDINAL

ORG) Academy ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

ORG ORG

First ORDINAL

2:1 CARDINAL

CARDINAL ORG

2:2 CARDINAL

CARDINAL ORG

Third ORDINAL

GPA ORG

13 CARDINAL

CARDINAL ORG

2:1 CARDINAL

CARDINAL ORG

ORG ORG

ORG ORG

30.5 CARDINAL

TIME ORG

hours TIME

TIME ORG

14 CARDINAL

TIME ORG

hours TIME

TIME ORG

weekly DATE

Mass (ORG) Communications ORG

ORG ORG

ORG ORG

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

22 DATE

QUANTITY ORG

hours TIME

QUANTITY ORG

GPA ORG

ORG ORG

Robert PERSON

ORG ORG

UK GPE

GPE ORG

GPA ORG

ORG ORG

Higher PERSON

ORG) ORG

ORG ORG

2012 DATE

2020 DATE

CARDINAL ORG

2009 DATE

Les PERSON

Ebdon PERSON

2 CARDINAL

years DATE

CNN ORG

ORG ORG

Sweden GPE

GPE ORG

Sweden GPE

GPE ORG

first ORDINAL

two CARDINAL

CARDINAL ORG

Sweden GPE

GPE ORG

GPE ORG

Egalia GPE

GPE ORG

Latin LANGUAGE

first ORDINAL

Egalia GPE

GPE ORG

10 CARDINAL

ORG ORG

15 CARDINAL

CARDINAL ORG

10 CARDINAL

14 CARDINAL

Sweden GPE

GPE ORG

fourth ORDINAL

1998 DATE

Lotta GPE

first ORDINAL

CARDINAL ORG

CARDINAL ORG

five CARDINAL

Stockholm GPE

GPE ORG

Rajalin GPE

ORG ORG

heroines CARDINAL

CARDINAL ORG

Bangladesh ORG

GPE ORG

Bangladesh GPE

GPE ORG

CARDINAL ORG

CARDINAL ORG

third ORDINAL

CARDINAL ORG

Shidhulai PERSON

TIME ORG

morning TIME

TIME ORG

CARDINAL ORG

CARDINAL ORG

30 CARDINAL

CARDINAL ORG

hundreds CARDINAL

CARDINAL ORG

CARDINAL ORG

UK GPE

GPE ORG

Sevenoaks (ORG) School (ORG ORG

UK GPE

GPE ORG

VR ORG

ORG ORG

three CARDINAL

CARDINAL ORG

CARDINAL ORG

Bali PERSON

Bali GPE

GPE ORG

three CARDINAL

CARDINAL ORG

CARDINAL ORG

10 CARDINAL

CARDINAL ORG

Ghana GPE

GPE ORG

Ghana PERSON

GPE ORG

TIME ORG

40 DATE

TIME ORG

hours TIME

TIME ORG

CARDINAL ORG

Canada GPE

GPE ORG

Canada GPE

GPE ORG

U.S. GPE

GPE ORG

iPads (ORG ORG

Google ORG

CARDINAL ORG

Netherlands GPE

GPE ORG

Steve PERSON

Apple ORG

iPad ORG

ORG ORG

Netherlands GPE

GPE ORG

Johannesburg GPE

GPE ORG

2016 DATE

CARDINAL ORG

Sweden GPE

GPE ORG

Egalia GPE

GPE ORG

Stockholm GPE

GPE ORG

Sweden GPE

GPE ORG

CARDINAL ORG

Singapore GPE

GPE ORG

Singapore GPE

GPE ORG

Pepper (ORG ORG

CARDINAL ORG

Denmark GPE

GPE ORG

Denmark GPE

GPE ORG

ORG ORG

ORG ORG

10 (PERCENT) percent PERCENT

Danish NORP

CARDINAL ORG

USA GPE

GPE ORG

GPE ORG

2018 DATE

19 CARDINAL

India GPE

GPE ORG

Botswana PERSON

GPE ORG

Japan GPE

GPE ORG

Spain GPE

GPE ORG

Rajalin PERSON

CNN ORG

ORG ORG

Lotta PERSON

Nicolaigarden ORG

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Egalia GPE

GPE ORG

Rajalin PERSON

two CARDINAL

CARDINAL ORG

Rajalin PERSON

Rajalin PERSON

Ben PERSON

Uppsala ORG

ORG) University ORG

Sweden GPE

GPE ORG

Oxford ORG

ORG) University ORG

England GPE

GPE ORG

Swedish NORP

years DATE

Swedish NORP

NORP PERSON

David PERSON

CNN ORG

ORG ORG

Rajalin PERSON

Rajalin PERSON

Eberhard PERSON

Sweden GPE

GPE ORG

Rajalin PERSON

Philip PERSON

Hwang PERSON

ORG ORG

Rajalin PERSON

CNN ORG

ORG ORG

Trump WORK OF ART

ORG ORG

ORG ORG

EU ORG

ORG ORG

one CARDINAL

nine CARDINAL

CARDINAL ORG

Britain GPE

GPE ORG

SUBSCRIBE PRODUCT

Invalid PERSON

TIME ORG

TIME ORG

six CARDINAL

years DATE

2010 DATE

CARDINAL ORG

2.2 CARDINAL

CARDINAL ORG

CARDINAL ORG

Britain GPE

GPE ORG

28.2 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

1.1 CARDINAL

CARDINAL ORG

CARDINAL ORG

Europe LOC

LOC ORG

3.4 CARDINAL

CARDINAL ORG

CARDINAL ORG

30.3 CARDINAL

CARDINAL ORG

CARDINAL ORG

2.2 CARDINAL

CARDINAL ORG

CARDINAL ORG

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1.2 CARDINAL

CARDINAL ORG

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EU ORG

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years DATE

Britain GPE

GPE ORG

EU ORG

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Theresa PERSON

British NORP

six CARDINAL

years DATE

EU ORG

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**Brexit PERSON** 

28 CARDINAL

CARDINAL ORG

European NORP

ORG ORG

Getty PERSON

29 CARDINAL

CARDINAL ORG

European NORP

ORG ORG

ORG ORG

11.2 PERSON

UK GPE

GPE ORG

2016 DATE

EU ORG

761,000 CARDINAL

CARDINAL ORG

508,000 CARDINAL

CARDINAL ORG

EU ORG

ORG ORG

669,000 CARDINAL

CARDINAL ORG

510,000 CARDINAL

CARDINAL ORG

EU ORG

ORG ORG

Eight CARDINAL

eight CARDINAL

CARDINAL ORG

European NORP

EU8 PERSON

FAC ORG

EU ORG

ORG ORG

2004 DATE

Czech NORP

GPE ORG

Estonia GPE

GPE ORG

Hungary GPE

GPE ORG

Latvia GPE

GPE ORG

Lithuania GPE

GPE ORG

Poland GPE

GPE ORG

Slovakia ORG

GPE ORG

Slovenia GPE

GPE ORG

EU ORG

ORG ORG

11 CARDINAL

Seven CARDINAL

eight CARDINAL

CARDINAL ORG

EU ORG

ORG ORG

CARDINAL ORG

eight CARDINAL

CARDINAL ORG

382,000 CARDINAL

CARDINAL ORG

EU ORG

ORG ORG

GPE ORG

Jane PERSON

EU ORG

ORG ORG

Brussels GPE

GPE ORG

CARDINAL ORG

701,000 CARDINAL

CARDINAL ORG

non - UK GPE

GPE ORG

EU ORG

ORG ORG

Yesterday DATE

CARDINAL ORG

CARDINAL ORG

thousands CARDINAL

CARDINAL ORG

European NORP

Britain GPE

GPE ORG

Theresa PERSON

UK GPE

GPE ORG

Homeless PERSON

24 (PERCENT) % PERCENT

597 CARDINAL

CARDINAL ORG

Three CARDINAL

CARDINAL ORG

four CARDINAL

CARDINAL ORG

today DATE

CARDINAL ORG

half CARDINAL

CARDINAL ORG

just two CARDINAL

CARDINAL ORG

165 CARDINAL

CARDINAL ORG

Gynaecologists (NORP ORG

today DATE

CARDINAL ORG

half CARDINAL

CARDINAL ORG

TIME ORG

minutes TIME

TIME ORG

CARDINAL ORG

CARDINAL ORG

hours TIME

TIME ORG

165 CARDINAL

CARDINAL ORG

NHS ORG

ORG ORG

82 CARDINAL

CARDINAL ORG

CARDINAL ORG

73 CARDINAL

CARDINAL ORG

TIME ORG

10 CARDINAL

TIME ORG

TIME ORG

TIME ORG

TIME ORG

TIME ORG

Another 43 CARDINAL

89 DATE

evening TIME

TIME ORG

England GPE

ORG ORG

Birmingham GPE

GPE ORG

Central ORG

Manchester PERSON

ORG) University ORG

Maureen PERSON

Treadwell PERSON

night TIME

TIME ORG

Jonathan PERSON

Labour ORG

five CARDINAL

NHS Jonathan PERSON

Labour ORG

night TIME

TIME ORG

Tories NORP

ORG ORG

n't GPE

five CARDINAL

ORG ORG

Elizabeth PERSON

**Duff PERSON** 

ORG ORG

2009 DATE

Imperial (ORG) College (ORG) ORG

ORG ORG

Cambridge GPE

ORG) University ORG

2010 DATE

overnight TIME

TIME ORG

Scottish NORP

a third CARDINAL

THE WEEK Angela EVENT

Owens PERSON

Ella PERSON

ORG ORG

today DATE

Angela PERSON

Owens PERSON

Ella PERSON

ORG ORG

today DATE

two CARDINAL

weeks DATE

Saturday DATE

afternoon TIME

TIME ORG

four CARDINAL

Christmas DATE

2013 DATE

31 DATE

Warrington ORG

ORG ORG

GP ORG

ORG ORG

Ella ORG

ORG ORG

11 CARDINAL

TIME ORG

TIME ORG

the Saturday DATE

two CARDINAL

CARDINAL ORG

Paul PERSON

Ella ORG

ORG ORG

CARDINAL ORG

ten CARDINAL

CARDINAL ORG

Warrington ORG

12 CARDINAL

English LANGUAGE

NHS ORG

ORG ORG

2014/15 CARDINAL

CARDINAL ORG

24 CARDINAL

1989 DATE

NHS ORG

ORG ORG

NHS ORG

ORG ORG

overnight TIME

TIME ORG

hours TIME

TIME ORG

Tomorrow DATE

NHS ORG

ORG ORG

first ORDINAL

Ofsted ORG

Biden PERSON

Russia GPE

GPE ORG

SolarWinds ORG

ORG ORG

Tesco ORG

ORG ORG

59,400 CARDINAL

CARDINAL ORG

CARDINAL ORG

119 CARDINAL

CARDINAL ORG

CARDINAL ORG

Tesco PERSON

ORG ORG

4 (PERCENT) % PERCENT

2015 DATE

CARDINAL ORG

CARDINAL ORG

Tesco ORG

ORG ORG

annual DATE

Tesco ORG

ORG ORG

Dave PERSON

today DATE

Tesco ORG

ORG ORG

UK GPE

```
GPE ORG
```

three CARDINAL

years DATE

UK GPE

GPE ORG

Tesco ORG

ORG ORG

2013 DATE

Tesco ORG

ORG ORG

2 (PERCENT) % PERCENT

0.5 CARDINAL

QUANTITY ORG

QUANTITY ORG

Lewis PERSON

Unilever ORG

ORG ORG

Tesco ORG

ORG ORG

2014 DATE

ORG ORG

Tesco ORG

ORG ORG

February DATE

2017 DATE

Tesco ORG

ORG ORG

years DATE

millions CARDINAL

CARDINAL ORG

Tesco PERSON

ORG ORG

Dave PERSON

Neil PERSON

Tesco ORG

ORG ORG

Asda ORG

Sainsbury ORG

Morrisons PERSON

fifth ORDINAL

Coca (ORG) - (ORG

Nestle ORG

ORG ORG

2025 DATE

ORG) Resources (ORG) Action ORG

1.9 CARDINAL

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

UK GPE

GPE ORG

Tesco PERSON

ORG ORG

Dr Richard PERSON

Wrap (ORG ORG

ORG ORG

UK GPE

GPE ORG

2025 DATE

200,000 CARDINAL

QUANTITY ORG

**QUANTITY ORG** 

UK GPE

GPE ORG

WRAP ORG

ORG ORG

27 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ONC

CARDINAL ORG

UK GPE

GPE ORG

2025 DATE

Ohio GPE

GPE ORG

163 CARDINAL

CARDINAL ORG

George PERSON

Soros PERSON

Soros PERSON

Craigslist NORP

Paid PERSON

6 DATE

TIME ORG

hours TIME

TIME ORG

30 MONEY

40 MONEY

70 DATE

neary \$ 1 MONEY

Soros PERSON

n't GPE

n't PERSON

Joe PERSON

George PERSON

Soros PERSON

an extra \$ 500 MONEY

Art PERSON

163 CARDINAL

CARDINAL ORG

Soros PERSON

Deluth GPE

GPE ORG

Charlie PERSON

Kirk PERSON

Turning PERSON

Turning PERSON

163 CARDINAL

CARDINAL ORG

Soros PERSON

NPR ORG

ORG ORG

NPR ORG

ORG ORG

NPR ORG

NPR ORG

ORG ORG

NPR ORG

NEK UKU

ORG ORG

months DATE

WHO ORG

WHO ORG

34 DATE

CARDINAL ORG

Monday DATE

October DATE

5th ORDINAL

n't PERSON

Michael PERSON

Ryan PERSON

WHO ORG

PERCENT ORG

10 (PERCENT) % PERCENT

CARDINAL ORG

35 DATE

CARDINAL ORG

CARDINAL ORG

Margaret PERSON

Harris PERSON

WHO ORG

ORG ORG

WHO ORG

ORG ORG

Ryan PERSON

CARDINAL ORG

7.8 CARDINAL

CARDINAL ORG

CARDINAL ORG

```
10 (PERCENT) % PERCENT
```

780 CARDINAL

CARDINAL ORG

CARDINAL ORG

1,061,539 CARDINAL

CARDINAL ORG

0.14 (PERCENT) % PERCENT

0.14 (PERCENT) % PERCENT

24 DATE

CARDINAL ORG

WHO ORG

3.4 (PERCENT) % PERCENT

March DATE

Covid PERSON

ORG ORG

Covid PERSON

Dr GPE

WHO ORG

ORG ORG

CNN ORG

ORG ORG

GPE ORG

ORG ORG

CNN ORG

ORG ORG

Jeff PERSON

ORG ORG

Friday DATE

US GPE

GPE ORG

US GPE

GPE ORG

CDC ORG

ORG ORG

US GPE

GPE ORG

CDC ORG

ORG ORG

Amazon ORG

ORG ORG

Orwell ORG

Eighty PERSON

Four CARDINAL

Trump ORG

ORG ORG

Kellyanne PERSON

Jezebel PERSON

Hannah PERSON

Arendt PERSON

1951 DATE

Donald PERSON

November DATE

Jews NORP

first ORDINAL

Nazi NORP

Trump ORG

ORG ORG

Jewish NORP

Europe LOC

LOC ORG

John PERSON

Naughton PERSON

David PERSON

Runciman PERSON

Talking PERSON

ORG ORG

n't GPE

20 CARDINAL

years DATE

Trump ORG

ORG ORG

Washington GPE

GPE ORG

Pennsylvania GPE

FAC ORG

March DATE

January DATE

Joshua PERSON

Lott PERSON

CARDINAL ORG

First ORDINAL

n't GPE

NORP PERSON

Second ORDINAL

Runciman PERSON

CARDINAL ORG

Arendt ORG

ORG ORG

Griselda PERSON

Arendt ORG

GPE ORG

two CARDINAL

CARDINAL ORG

n't ORG

n't GPE

Arendt ORG

ORG ORG

Germany GPE

GPE ORG

1906 DATE

1933 DATE

Palestine GPE

GPE ORG

Jewish NORP

Hitler PERSON

1964 DATE

n't PERSON

Nazis NORP

Germans NORP

1933 DATE

Jewish NORP

Arendt ORG

Jewish NORP

CARDINAL ORG

CARDINAL ORG

Paris GPE

GPE ORG

German NORP

NORP PERSON

1937 DATE

GPE ORG

1941 DATE

FAC ORG

France GPE

GPE ORG

Holocaust EVENT

Auschwitz PERSON

ORG ORG

n't GPE

Arendt (ORG ORG

GPE ORG

1972 DATE

ORG ORG

1960s DATE

Eichmann PERSON

NORP PERSON

Jerusalem GPE

GPE ORG

Adolf PERSON

Jews NORP

Jews NORP

three CARDINAL

TIME ORG

minutes TIME

TIME ORG

Martin PERSON

Heidegger PERSON

20s DATE

Nazi NORP

Nazi NORP

US GPE

today DATE

ORG ORG

n't GPE

Arendt (ORG) 's ORG

Origins PERSON

1958 DATE

Holocaust EVENT

Arendt PERSON

ORG ORG

two CARDINAL

CARDINAL ORG

First ORDINAL

Greek NORP

Greek NORP

Arendt ORG

CARDINAL ORG

Trump ORG

ORG ORG

Leninist PERSON

ORG ORG

Trump ORG

ORG ORG

Mark PERSON

Davis PERSON

Bauman PERSON

Leeds PERSON

GPE ORG

1970 DATE

Downing (FAC) Street ORG

FAC ORG

January DATE

Ben PERSON

CARDINAL ORG

First ORDINAL

Muslims NORP

today DATE

Hannah PERSON

Arendt PERSON

Downing (FAC) Street ORG

FAC ORG

Davis PERSON

CARDINAL ORG

CARDINAL ORG

n't GPE

Arendt PERSON

ORG ORG

ORG ORG

The (WORK\_OF\_ART) Twilight WORK\_OF\_ART

NASA ORG

ORG ORG

Antarctica LOC

LOC ORG

1960s DATE

sci (ORG) - ( ORG

ORG ORG

ORG ORG

GPE ORG

ORG ORG

NASA ORG

Antenna PERSON

ANITA ORG

ORG ORG

Antarctica LOC

LOC ORG

Earth LOC

LOC ORG

Earth LOC

LOC ORG

Manoa PERSON

GPE ORG

ANITA ORG

ORG ORG

Earth LOC

LOC ORG

ORG ORG

ANITA PERSON

ORG ORG

Peter PERSON

Gorham PERSON

ORG ORG

Earth LOC

LOC ORG

Gorham PERSON

Cornell GPE

ORG) University ORG

ORG ORG

New (ORG) Scientist ORG

ORG ORG

ORG ORG

13.8 CARDINAL

years DATE

two CARDINAL

CARDINAL ORG

Ibrahim PERSON

Christian (NORP ORG

One CARDINAL

Catholic NORP

Kentucky GPE

GPE ORG

Native NORP

Vietnam GPE

GPE ORG

Saturday DATE

Covington PERSON

Park Hillswere FAC

Washington GPE

GPE ORG

Friday DATE

CARDINAL ORG

Phillips ORG

ORG ORG

Donald PERSON

MAGA ORG

ORG ORG

Phillips PERSON

ORG ORG

Friday DATE

Alyssa PERSON

Chris PERSON

**Evans PERSON** 

Diocese NORP

Covington ORG

Nathan (ORG) ORG

Native (NORP ORG

Americans NORP

Phillips PERSON

the Church (ORG) 's ORG

Covington PERSON

ORG ORG

Joe PERSON

Democrat NORP

64 CARDINAL

Phillips PERSON

ORG ORG

Nebraska GPE

GPE ORG

Omaha GPE

GPE ORG

ORG ORG

Washington GPE

Shoshana PERSON

four CARDINAL

years DATE

Phillips PERSON

ORG ORG

Phillips PERSON

ORG ORG

annual DATE

Arlington GPE

Native PERSON

American NORP

Native (NORP) ORG

American NORP

the Standing (ORG) Rock (ORG ORG

2016 DATE 2017 DATE

North (GPE) Dakota ORG

GPE ORG

Democratic NORP

US GPE

GPE ORG

New (GPE) Mexico ORG

GPE ORG

Twitter PERSON

Phillips PERSON

ORG ORG

Reuters ORG

ORG ORG

Associated ORG

ORG ORG

Garba PERSON

Britain GPE

GPE ORG

0.7 PERSON

an extra MONEY

two CARDINAL

years DATE

European NORP

ORG ORG

yesterday DATE

ORG) Department ORG

ORG ORG

Britain GPE

GPE ORG

EU ORG

ORG ORG

UK GPE

GPE ORG

second ORDINAL

yesterday DATE

Britain GPE

GPE ORG

0.7 PERSON

1bn MONEY

two CARDINAL

years DATE

Europe LOC

LOC ORG

Jonathan PERSON

```
Labour ORG
today DATE
3 CARDINAL
4 CARDINAL
months DATE
Labour ORG
Britain GPE
GPE ORG
3 CARDINAL
4 CARDINAL
months DATE
ORG ORG
Labour ORG
National (ORG) Child (ORG) Health (ORG) Fund ORG
ORG ORG
Britain GPE
GPE ORG
Today DATE
Labour ORG
Jonathan PERSON
the national Health (ORG) Visitor ORG
2015 DATE
CARDINAL ORG
CARDINAL ORG
2017/18 CARDINAL
CARDINAL ORG
5 CARDINAL
ORG ORG
8,244 CARDINAL
CARDINAL ORG
NHS ORG
ORG ORG
December DATE
2017 DATE
August DATE
2013 DATE
10,309 CARDINAL
CARDINAL ORG
October DATE
2015 DATE
CARDINAL ORG
CARDINAL ORG
2,065 CARDINAL
CARDINAL ORG
20 (PERCENT) % PERCENT
two CARDINAL
vears DATE
12 DATE
17 (PERCENT) % PERCENT
LOC ORG
```

LOC ORG

LOC ORG

England GPE

GPE ORG

17 DATE

6 CARDINAL

8 CARDINAL

London GPE

GPE ORG

1/3 CARDINAL

CARDINAL ORG

12 CARDINAL

CARDINAL ORG

London GPE

GPE ORG

44 DATE

North ORG

LOC ORG

LOC ORG

two CARDINAL

half CARDINAL

PERCENT ORG

64 (PERCENT) % PERCENT

London GPE

GPE ORG

73 (PERCENT) % PERCENT

LOC ORG

LOC ORG

LOC ORG

England GPE

LOC ORG

Jonathan PERSON

today DATE

Infant PERSON

first ORDINAL

ORG ORG

ORG ORG

UNICEF ORG

ORG ORG

Scotland GPE

GPE ORG

Northern (GPE) Ireland ORG

GPE ORG

70 (PERCENT) % PERCENT

2030 DATE

Jonathan PERSON

Labour ORG

London GPE

GPE ORG

Michael PERSON

```
UK GPE
GPE ORG
NHS ORG
ORG ORG
David PERSON
Cameron PERSON
Theresa PERSON
20 (PERCENT) percent PERCENT
two CARDINAL
years DATE
England GPE
GPE ORG
UK GPE
GPE ORG
Health (ORG) Visitors ORG
ORG ORG
3 CARDINAL
4 CARDINAL
months DATE
an extra £ MONEY
National (ORG) Child (ORG) Health (ORG) Fund ORG
ORG ORG
20 CARDINAL
100 CARDINAL
Rick PERSON
Scott PERSON
Democrats NORP
Liberal (GPE) States ORG
GPE ORG
Cambodia GPE
GPE ORG
Sam PERSON
GPE ORG
seven CARDINAL
CARDINAL ORG
CNRP ORG
ORG ORG
Cambodia GPE
GPE ORG
Sam PERSON
GPE ORG
seven CARDINAL
CARDINAL ORG
CNRP ORG
ORG ORG
ORG ORG
Sam PERSON
eight CARDINAL
CARDINAL ORG
```

Chin PERSON

yesterday DATE

**Duch PERSON** 

Monday DATE

Rainsy PERSON

25 CARDINAL

years DATE

Mu PERSON

Sochua PERSON

Eng PERSON

22 CARDINAL

years DATE

six CARDINAL

CARDINAL ORG

20 CARDINAL

years DATE

Rainsy PERSON

Tioulong PERSON

Saumura GPE

Ou PERSON

Vann PERSON

Long Ry GPE

GPE ORG

Nuth PERSON

ORG ORG

the Kingdom (GPE GPE

November DATE

9 CARDINAL

2019 DATE

Rainsy PERSON

Cambodian NORP

Hun PERSON

Cambodian NORP

Rainsy PERSON

Hun PERSON

Cambodia GPE

GPE ORG

ORG ORG

Malin PERSON

Malin PERSON

CNRP ORG

ORG ORG

Cambodia GPE

GPE ORG

130 CARDINAL

CARDINAL ORG

nine CARDINAL

CARDINAL ORG

CNRP ORG

US GPE

ORG ORG

Phnom GPE

GPE ORG

yesterday DATE

nine CARDINAL

CARDINAL ORG

Cambodian NORP

American NORP

CARDINAL ORG

CARDINAL ORG

130 CARDINAL

CARDINAL ORG

Cambodia GPE

GPE ORG

2022 DATE

Cambodian NORP

ORG ORG

Sok PERSON

yesterday DATE

ex - CNRP ORG

ORG ORG

2022 DATE

2023 DATE

CARDINAL ORG

ORG ORG

America GPE

GPE ORG

Christian NORP

Americans NORP

Christianity NORP

Nazi NORP

Germany GPE

GPE ORG

GPE ORG

Donald PERSON

66 CARDINAL

CARDINAL ORG

GPE ORG

GPE) America ORG

GPE ORG

IRS ORG

ORG ORG

years DATE

Americans NORP

America GPE

GPE ORG

n't GPE

Trump ORG

ORG) Church ORG

Kafka PERSON

Kafka PERSON

Kafka PERSON

thirty CARDINAL

years DATE

Today DATE

hundreds CARDINAL

two CARDINAL

CARDINAL ORG

1988 DATE

thirty DATE

years DATE

two CARDINAL

Netflix GPE

GPE ORG

CARDINAL ORG

Republican NORP

ORG ORG

CARDINAL ORG

CARDINAL ORG

two CARDINAL

CARDINAL ORG

degrees Celsius PERSON

CARDINAL ORG

CARDINAL ORG

two CARDINAL

CARDINAL ORG

three CARDINAL

decades DATE

CARDINAL ORG

three CARDINAL

decades DATE

GPE ORG

Scientific ORG

ten CARDINAL

CARDINAL ORG

CARDINAL ORG

two CARDINAL

QUANTITY ORG

QUANTITY ORG

CARDINAL ORG

CARDINAL ORG

two CARDINAL

CARDINAL ORG

first ORDINAL

Nature GPE

GPE ORG

thousands CARDINAL

CARDINAL ORG

GPE ORG

Texans NORP

Kafkaesque NORP

European NORP

ORG ORG

Indonesia GPE

GPE ORG

American NORP

China GPE

GPE ORG

COVID-19 ORG

ORG ORG

GPE ORG

China GPE

GPE ORG

CARDINAL ORG

82,000 CARDINAL

CARDINAL ORG

CARDINAL ORG

3,300 CARDINAL

CARDINAL ORG

China GPE

GPE ORG

Wuhan GPE

GPE ORG

11 CARDINAL

CARDINAL ORG

CARDINAL ORG

GPE ORG

CARDINAL ORG

500,000 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

18,000 CARDINAL

CARDINAL ORG

China GPE

GPE ORG

China GPE

GPE ORG

CARDINAL ORG

21 DATE

CARDINAL ORG

CARDINAL ORG

January DATE

February DATE

China GPE

ORG) Ltd. ORG

ORG ORG

CARDINAL ORG

```
7.25 CARDINAL
```

CARDINAL ORG

China GPE

GPE ORG

Unicom Hong PERSON

ORG ORG

7.8 CARDINAL

CARDINAL ORG

CARDINAL ORG

January DATE

February DATE

China GPE

ORG) Corp. ORG

ORG ORG

5.6 CARDINAL

CARDINAL ORG

CARDINAL ORG

February DATE

ORG ORG

China GPE

GPE ORG

Tang PERSON

China GPE

GPE ORG

ORG ORG

China GPE

GPE ORG

Samantha PERSON

Hoffman PERSON

Australian NORP

Hoffman PERSON

ORG ORG

n't GPE

Lunar PERSON

Bloomberg PERSON

Hoffman PERSON

China GPE

GPE ORG

U.S. GPE

GPE ORG

China GPE

GPE ORG

ORG ORG

Fortune (ORG) Magazine (ORG WORK\_OF\_ART

Hoffman PERSON

China GPE

GPE ORG

China GPE

GPE ORG

Hoffman PERSON

GPE ORG

Chinese NORP

Business (ORG) Insider (ORG ORG

Beijing GPE

GPE ORG

Hoffman PERSON

China GPE

GPE ORG

Chinese NORP

Communist NORP

ORG) Party ORG

ORG ORG

Hoffman PERSON

Hoffman PERSON

China GPE

GPE ORG

CARDINAL ORG

China GPE

GPE ORG

Hoffman PERSON

U.S. GPE

GPE ORG

U.S. GPE

GPE ORG

State ORG

ORG ORG

Michael PERSON

China GPE

GPE ORG

Mike PERSON

CNN ORG

ORG ORG

America GPE

GPE ORG

Hoffman PERSON

China GPE

GPE ORG

Hoffman PERSON

Chinese NORP

Communist NORP

ORG) Party ORG

ORG ORG

21 DATE

CARDINAL ORG

CARDINAL ORG

China GPE

GPE ORG

Coronavirus ORG

GPE ORG

ORG ORG

Alaska GPE

GPE ORG

Arctic LOC

LOC ORG

Woods ORG

**OUANTITY ORG** 

QUANTITY ORG

QUANTITY ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

hundreds CARDINAL

CARDINAL ORG

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

1.7 CARDINAL

QUANTITY ORG

QUANTITY ORG

Alaska GPE

GPE ORG

350 CARDINAL

QUANTITY ORG

miles PERSON

QUANTITY ORG

LOC ORG

Arctic LOC

LOC ORG

Circle PERSON

LOC ORG

two CARDINAL

CARDINAL ORG

the Yukon (ORG ORG

Kuskokwim PERSON

three CARDINAL

QUANTITY ORG

**OUANTITY ORG** 

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

Max PERSON

Holmes PERSON

Alaska GPE

GPE ORG

Sarah PERSON

Ludwig PERSON

ORG ORG

Laura PERSON

Jardine PERSON

Jardine PERSON

John PERSON

July DATE

Woods ORG

LOC ORG

LOC ORG

13 CARDINAL

CARDINAL ORG

60 CARDINAL

QUANTITY ORG

miles PERSON

**QUANTITY ORG** 

Bethel (ORG ORG

6,000 CARDINAL

CARDINAL ORG

Woods ( ORG

FAC ORG

Sue PERSON

Woods ORG

LOC ORG

LOC ORG

Alaska GPE

GPE ORG

QUANTITY ORG

QUANTITY ORG

2,100 CARDINAL

QUANTITY ORG

QUANTITY ORG

Vladimir PERSON

Romanovsky PERSON

the (ORG) University ORG

ORG ORG

Alaska GPE

ORG) Fairbanks ORG

ORG ORG

65 CARDINAL

QUANTITY ORG

**OUANTITY ORG** 

3 CARDINAL

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

5.5 CARDINAL

QUANTITY ORG

QUANTITY ORG

GPE ORG

decades DATE

CARDINAL ORG

QUANTITY ORG

8 CARDINAL

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

LAW ORG

3 CARDINAL

LAW ORG

zero CARDINAL

CARDINAL ORG

Romanovsky PERSON

Alaska GPE

GPE ORG

Romanovsky PERSON

Bethel (GPE ORG

QUANTITY ORG

4 CARDINAL

**QUANTITY ORG** 

QUANTITY ORG

Bethel GPE

LOC ORG

winter DATE

spring DATE

years DATE

Sarah PERSON

Ludwig PERSON

ORG ORG

Laura PERSON

Jardine PERSON

Jardine PERSON

John PERSON

Sarah PERSON

Ludwig PERSON

ORG ORG

Laura PERSON

Jardine PERSON

Jardine PERSON

John PERSON

1970s DATE

summer DATE

2015 DATE

Bethel (GPE ORG

Natali PERSON

CARDINAL ORG

QUANTITY ORG

1.5 CARDINAL

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

a year DATE

GPE ORG

Alaskan NORP

Alaska GPE

GPE ORG

Holmes PERSON

Siberia LOC

LOC ORG

Canada GPE

GPE ORG

CARDINAL ORG

thousands CARDINAL

CARDINAL ORG

CARDINAL ORG

thousands CARDINAL

CARDINAL ORG

years DATE

Correction (EVENT ORG

August DATE

23 CARDINAL

2017 DATE

1.5 CARDINAL

QUANTITY ORG

QUANTITY ORG

a year DATE

1.5 CARDINAL

QUANTITY ORG

**QUANTITY ORG** 

**QUANTITY ORG** 

August DATE

25 CARDINAL

2017 DATE

CARDINAL ORG

the (ORG) University ORG

ORG ORG

Alaska GPE

ORG) Fairbanks ORG

ORG ORG

Vladimir PERSON

Romanovsky PERSON

Ramonovsky PERSON

Alaska PERSON

GPE ORG

Horowitz PERSON

Redistricting PERSON

Berger PERSON

NC GPE

GPE ORG

Muhammadu PERSON

Buhari PERSON

Monday DATE

Lagos GPE

GPE ORG

Ogun PERSON

GPE ORG

Abuja (GPE ORG

14 DATE

Monday DATE

TIME ORG

Nigerians NORP

Nigerians NORP

BUHARI GPE

GPE ORG

GPE ORG

GPE ORG

GPE ORG

CARDINAL ORG

ORG ORG

HOUSE ORG

ORG ORG

MONDAY DATE

APRIL DATE

2020 DATE

1 CARDINAL

Fellow PERSON

Nigerians NORP

ORG ORG

2 CARDINAL

Sunday DATE

29th DATE

March DATE

2020 DATE

Lagos GPE

GPE ORG

Ogun PERSON

GPE ORG

ORG ORG

fourteen DATE

Monday DATE

March DATE

2020 DATE

3 CARDINAL

CARDINAL ORG

State ORG

ORG ORG

4 CARDINAL

CARDINAL ORG

CARDINAL ORG

5 CARDINAL

CARDINAL ORG

CARDINAL ORG

COVID-19 ORG

ORG ORG

6 CARDINAL

CARDINAL ORG

Coronavirus LOC

LOC ORG

COVID-19 PERSON

ORG ORG

ORG) Nigeria ORG

ORG ORG

Nigerian NORP

ORG ORG

**NSCIA PERSON** 

ORG ORG

9 CARDINAL

CARDINAL ORG

10 CARDINAL

CARDINAL ORG

11 CARDINAL

CARDINAL ORG

14 DATE

12 CARDINAL

CARDINAL ORG

13 CARDINAL

CARDINAL ORG

92 (PERCENT) % PERCENT

1,500 CARDINAL

CARDINAL ORG

over 7,000 CARDINAL

CARDINAL ORG

Healthcare ORG

ORG ORG

NCDC ORG

ORG ORG

19 CARDINAL

CARDINAL ORG

15 CARDINAL

CARDINAL ORG

Abuja (GPE ORG

today DATE

CARDINAL ORG

1,000 CARDINAL

CARDINAL ORG

16 CARDINAL

CARDINAL ORG

CARDINAL ORG

State ORG

17 CARDINAL

CARDINAL ORG

weeks DATE

19 CARDINAL

CARDINAL ORG

Lagos GPE

GPE ORG

Ogun PERSON

GPE ORG

ORG ORG

20 CARDINAL

CARDINAL ORG

CARDINAL ORG

weeks DATE

21 CARDINAL

CARDINAL ORG

COVID-19 ORG

ORG ORG

22 CARDINAL

CARDINAL ORG

Nigeria GPE

GPE ORG

23 CARDINAL

CARDINAL ORG

30th ORDINAL

March DATE

2020 DATE

CARDINAL ORG

780,000 CARDINAL

CARDINAL ORG

24 CARDINAL

CARDINAL ORG

Yesterday DATE

COVID-19 PERSON

CARDINAL ORG

CARDINAL ORG

eight CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

fifty CARDINAL

CARDINAL ORG

CARDINAL ORG

two CARDINAL

weeks DATE

25 CARDINAL

CARDINAL ORG

fourteen DATE

CARDINAL ORG

70,000 CARDINAL

CARDINAL ORG

27 CARDINAL

CARDINAL ORG

Nigeria GPE

GPE ORG

131 CARDINAL

CARDINAL ORG

COVID-19 ORG

ORG ORG

12 CARDINAL

CARDINAL ORG

30th DATE

March DATE

2020 DATE

two CARDINAL

CARDINAL ORG

28 CARDINAL

CARDINAL ORG

TIME ORG

morning TIME

TIME ORG

Nigeria PERSON

GPE ORG

323 CARDINAL

CARDINAL ORG

twenty CARDINAL

CARDINAL ORG

CARDINAL ORG

GPE ORG

GPE ORG

54 (PERCENT) % PERCENT

Nigeria GPE

GPE ORG

FCT ORG

ORG ORG

two CARDINAL

CARDINAL ORG

71 (PERCENT) % PERCENT

Nigeria GPE

GPE ORG

29 CARDINAL

CARDINAL ORG

two CARDINAL

CARDINAL ORG

30 CARDINAL

CARDINAL ORG

Lagos GPE

GPE ORG

FCT ORG

States GPE

GPE ORG

32 CARDINAL

CARDINAL ORG

33 CARDINAL

CARDINAL ORG

Nigerians NORP

34 CARDINAL

CARDINAL ORG

CARDINAL ORG

COVID-19 PERSON

ORG ORG

35 CARDINAL

CARDINAL ORG

COVID-19 ORG

ORG ORG

Today DATE

37 CARDINAL

CARDINAL ORG

2 CARDINAL

CARDINAL ORG

Nigerians NORP

second ORDINAL

38 CARDINAL

CARDINAL ORG

CARDINAL ORG

Report ORG

ORG ORG

ORG ORG

Lagos GPE

GPE ORG

Ogun PERSON

GPE ORG

FCT ORG

ORG ORG

14 DATE

11:59 TIME

TIME ORG

TIME ORG

Monday DATE

April DATE

2020 DATE

39 CARDINAL

CARDINAL ORG

Mosques PERSON

FAC ORG

Makkah PERSON

Madina GPE

GPE ORG

St. ORG

Peter PERSON

GPE ORG

GPE ORG

Notre GPE

FAC ORG

Paris GPE

GPE ORG

Easter (FAC ORG

CARDINAL ORG

CARDINAL ORG

10 CARDINAL

CARDINAL ORG

India GPE

GPE ORG

Italy GPE

GPE ORG

France GPE

GPE ORG

40 CARDINAL

CARDINAL ORG

41 CARDINAL

CARDINAL ORG

42 CARDINAL

CARDINAL ORG

45 CARDINAL

CARDINAL ORG

46 CARDINAL

CARDINAL ORG

47 CARDINAL

CARDINAL ORG

CARDINAL ORG

two CARDINAL

weeks DATE 48 CARDINAL

CARDINAL ORG

2.6 CARDINAL

CARDINAL ORG

CARDINAL ORG

3.6 CARDINAL

CARDINAL ORG

CARDINAL ORG

two CARDINAL

weeks DATE

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

49 CARDINAL

Fellow PERSON

Nigerians NORP

51 CARDINAL

CARDINAL ORG

State ORG

ORG ORG

ORG ORG

52 CARDINAL

CARDINAL ORG

weeks DATE

ORG ORG

ORG ORG

54 CARDINAL

CARDINAL ORG

55 CARDINAL

CARDINAL ORG

ORG ORG

Digital ORG

ORG) Technology ORG

Transportation ORG

Interior ORG

Health (ORG ORG

ORG ORG

Nigerian NORP

COVID-19 ORG

ORG ORG

56 CARDINAL

CARDINAL ORG

National (ORG) Food (ORG) Security ORG

ORG ORG

ORG ORG

COVID-19 ORG

ORG ORG

our 2020 DATE

season DATE

ORG ORG

COVID-19 ORG

ORG ORG

Nigerians NORP

GPE ORG

GPE) Nigeria ORG

GPE ORG

COVID-19 PERSON

COVID-19 ORG

ORG ORG

99.99 CARDINAL

Plexiglas PERSON

American NORP

CARDINAL ORG

U.S. GPE

GPE ORG

Anthony PERSON

Chicago GPE

GPE ORG

YouTube ORG

ORG ORG

Twitter PERSON

CARDINAL ORG

CARDINAL ORG

26,000 CARDINAL

CARDINAL ORG

YouTube ORG

ORG ORG

Pennsylvania GPE

GPE ORG

Tom PERSON

Wolf PERSON

Achieve PERSON

Oxford (ORG) University ORG

ORG ORG

six CARDINAL

CARDINAL ORG

ILI ORG

LOC ORG

Oxford ( ORG

two CARDINAL

decades DATE

CARDINAL ORG

CARDINAL ORG

1,600 CARDINAL

CARDINAL ORG

3 (PERCENT) percent PERCENT

ORG ORG

Cambridge GPE

ORG) University ORG

ORG ORG

Oxford ORG

ORG ORG

Influenza (ORG) Journal ORG

ORG ORG

two CARDINAL

CARDINAL ORG

two CARDINAL

CARDINAL ORG

**Graphs PERSON** 

Austria PERSON

GPE ORG

CARDINAL ORG

first ORDINAL

```
10 CARDINAL
```

four CARDINAL

CARDINAL ORG

Austria GPE

GPE ORG

Germany GPE

GPE ORG

French NORP

1,000 (PERCENT) percent PERCENT

daily DATE

CARDINAL ORG

Spain GPE

GPE ORG

French NORP

Spain GPE

GPE ORG

zero CARDINAL

CARDINAL ORG

Europe LOC

LOC ORG

Spain GPE

GPE ORG

1,500 (PERCENT) percent PERCENT

three CARDINAL

months DATE

GPE ORG

1,500 (PERCENT) percent PERCENT

Europe LOC

LOC ORG

Belgium GPE

GPE ORG

British NORP

Italy GPE

GPE ORG

Ireland GPE

GPE ORG

Portugal GPE

GPE ORG

Israel GPE

GPE ORG

GPE ORG

Americans NORP

Germans NORP

Europe LOC

LOC ORG

U.S. GPE

GPE ORG

California GPE

GPE ORG

June DATE

```
300 (PERCENT) percent PERCENT
four CARDINAL
months DATE
Hawaii GPE
GPE ORG
CARDINAL ORG
U.S. GPE
GPE ORG
PERCENT ORG
1,000 (PERCENT) percent PERCENT
Texas GPE
GPE ORG
Georgia GPE
GPE ORG
n't PERSON
ORG ORG
CDC ORG
ORG ORG
Sweden GPE
GPE ORG
Georgia GPE
GPE ORG
n't GPE
COVID-19 ORG
ORG ORG
Nobel WORK_OF_ART
COVID ORG
Martin PERSON
Kulldorff PERSON
Harvard ORG
COVID ORG
ORG ORG
300 CARDINAL
years DATE
12 CARDINAL
CARDINAL ORG
Graphs PERSON
Mask PERSON
AdvertisementsOne ORG
February DATE
3 CARDINAL
2020 DATE
Iowa GPE
GPE ORG
first ORDINAL
fifty CARDINAL
CARDINAL ORG
American NORP
Twitter PERSON
TIME ORG
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TIME ORG

TIME ORG

Iowa GPE

GPE ORG

CARDINAL ORG

first ORDINAL

Pete PERSON

Iowa GPE

GPE ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

Pete PERSON

Atlantic LOC

LOC ODC

LOC ORG

February DATE

3 CARDINAL

2020 DATE

**BURLINGTON ORG** 

GPE ORG

Iowa GPE

GPE ORG

Democratic NORP

Iowans NORP

hours TIME

TIME ORG

Joe PERSON

Biden PERSON

Elizabeth PERSON

Warren PERSON

Bernie PERSON

Dockendorff PERSON

tonight TIME

TIME ORG

62 CARDINAL

Danville PERSON

ORG ORG

January DATE

2020 DATE

Democrats NORP

Iowa GPE

GPE ORG

first ORDINAL

Dockendorff PERSON

Iowa GPE

GPE ORG

Eastern ORG

ORG ORG

ORG ORG

Twitter PERSON

Iowa GPE

GPE ORG

CARDINAL ORG

1 CARDINAL

CARDINAL ORG

Robby PERSON

Mook PERSON

Hillary PERSON

Clinton PERSON

2016 DATE

Iowa GPE

GPE ORG

Mook PERSON

Monday DATE

TIME ORG

TIME ORG

Iowa GPE

GPE ORG

Shadow PRODUCT

Mook PERSON

January DATE

29 CARDINAL

2020 DATE

Mook PERSON

Iowa GPE

GPE ORG

2018 DATE

Harvard ORG

night TIME

TIME ORG

Robby PERSON

Mook PERSON

Matt PERSON

Mook PERSON

2016 DATE

Hillary PERSON

Clinton PERSON

Rhodes PERSON

GPE ORG

Mitt PERSON

Romney PERSON

2012 DATE

2018 DATE

Russian NORP

Clinton PERSON

DNC ORG

ORG ORG

2016 DATE

Mook PERSON

February DATE

4 CARDINAL

2020 DATE

Des Moines Register PERSON

Clinton PERSON

Iowa GPE

GPE ORG

Nevada GPE

GPE ORG

February DATE

3 CARDINAL

2020 DATE

Iowa GPE

GPE ORG

Iowa GPE

GPE ORG

Democrats NORP

Hillary PERSON

Clinton PERSON

2016 DATE

Nevada GPE

ORG ORG

February DATE

4 CARDINAL

2020 DATE

Feb. DATE

22 DATE

2020 DATE

CARDINAL ORG

Jimmy PERSON

Hickey PERSON

Shadow Inc. ORG

ORG ORG

Tuesday DATE

Gerard PERSON

Krista PERSON

Davis PERSON

Clinton PERSON

2016 DATE

Shadow PRODUCT

Iowa GPE

GPE ORG

DNC ORG

Bernie PERSON

ORG ORG

2016 DATE

Bernie PERSON

2016 DATE

Iowa GPE

GPE ORG

2020 DATE

February DATE

3 CARDINAL

2020 DATE

ORG ORG

Democrats NORP

DNC ORG

ORG ORG

Bernie PERSON

Hillary PERSON

Clinton PERSON

the 2016 DATE

Iowa GPE

GPE ORG

Nevada GPE

GPE ORG

Feb. DATE

22 DATE

three CARDINAL

CARDINAL ORG

two CARDINAL

CARDINAL ORG

Iowa GPE

GPE ORG

Sanders ORG

ORG ORG

2016 DATE

CARDINAL ORG

December DATE

2017 DATE

ORG ORG

2020 DATE

**Iowans NORP** 

2020 DATE

Iowa GPE

GPE ORG

Democratic NORP

decades DATE

2020 DATE

ORG ORG

ORG ORG

December DATE

2017 DATE

Iowa GPE

GPE ORG

five CARDINAL

decades DATE

DNC ORG

ORG ORG

Tom PERSON

Perez PERSON

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2016 DATE
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Jeff PERSON

Jeff PERSON

2016 DATE

CARDINAL ORG

Martin PERSON

Iowa GPE

GPE ORG

0.6 (PERCENT) percent PERCENT

Iowa GPE

GPE ORG

CARDINAL ORG

CARDINAL ORG

two CARDINAL

CARDINAL ORG

DNC ORG

ORG ORG

2016 DATE

CARDINAL ORG

Bernie PERSON

Hillary PERSON

Clinton PERSON

Sanders ORG

ORG ORG

Weaver ORG

ORG ORG

DNC ORG

ORG ORG

2020 DATE

Iowa GPE

GPE ORG

Iowa GPE

GPE ORG

Rumor # ( ORG

CARDINAL ORG

3 CARDINAL

CARDINAL ORG

Iowa GPE

GPE ORG

Democratic NORP

ORG ORG

50 (PERCENT) percent PERCENT

5PM ORG

CARDINAL ORG

February DATE

4 CARDINAL

2020 DATE

ORG ORG

TIME ORG

morning TIME

TIME ORG

Iowa GPE

GPE ORG

25 (PERCENT) % PERCENT

ORG ORG

ORG ORG

three CARDINAL

CARDINAL ORG

1:14 TIME

TIME ORG

TIME ORG

TIME ORG

February DATE

4 CARDINAL

2020 DATE

NPR ORG

ORG ORG

Tuesday DATE

afternoon TIME

TIME ORG

February DATE

4 CARDINAL

2020 DATE

Democratic NORP

ORG ORG

5 CARDINAL

TIME ORG

TIME ORG

TIME ORG

February DATE

4 CARDINAL

2020 DATE

Tuesday DATE

February DATE

4 CARDINAL

2020 DATE

Troy PERSON

ORG ORG

NPR ORG

ORG ORG

Holly Christine PERSON

January DATE

2020 DATE

Asian NORP

ORG ORG

first ORDINAL

TIME ORG

evening TIME

TIME ORG

January DATE

31 DATE

2020 DATE

NPR ORG

Martin PERSON

February DATE

3 CARDINAL

2020 DATE

Iowa GPE

GPE ORG

Elesha GPE

Scott PERSON

GPE ORG

Democratic NORP

PIN ORG

ORG ORG

n't GPE

GPE) County Democratic Party Co - ORG

Tom PERSON

n't GPE

Courtney PERSON

TIME ORG

hours TIME

TIME ORG

NPR ORG

ORG ORG

Nevada GPE

ORG ORG

February DATE

22 DATE

2020 DATE

Iowa GPE

GPE ORG

TIME ORG

TIME ORG

Nevada GPE

GPE ORG

February DATE

2020 DATE

CARDINAL ORG

Iowa GPE

GPE ORG

2016 DATE

Iowa GPE

GPE ORG

Iowa GPE

GPE ORG

Twitter PERSON

2016 DATE

Iowa GPE

GPE ORG

Shadow PRODUCT

Iowa GPE

GPE ORG

Democratic NORP

ORG ORG

Iowa GPE

GPE ORG

2020 DATE

the day DATE

Invalid ORG

Email Something PERSON

Yesterday DATE

UK GPE

GPE ORG

Fathers (ORG) 4 ORG

ORG ORG

GPE ORG

ComRes ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

1000 CARDINAL

CARDINAL ORG

37 DATE

UK GPE

GPE ORG

2.3 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

five CARDINAL

CARDINAL ORG

20 CARDINAL

22 DATE

CARDINAL ORG

third ORDINAL

CARDINAL ORG

32 DATE

CARDINAL ORG

Matt PERSON

Fathers (ORG) 4 ORG

ORG ORG

Telegraph ORG

ORG ORG

millions CARDINAL

CARDINAL ORG

Women (ORG) 's ORG

ComRes ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

six CARDINAL

CARDINAL ORG

17 DATE

a quarter DATE

27 DATE

2001 DATE

750,000 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

every 15 minutes TIME

four CARDINAL

CARDINAL ORG

10 CARDINAL

CARDINAL ORG

CARDINAL ORG

five CARDINAL

CARDINAL ORG

UK GPE

GPE ORG

first ORDINAL

Today DATE

Nicola ORG

ORG ORG

Dandridge PERSON

CBE PERSON

ORG ORG

Chris PERSON

Dame Julia PERSON

first ORDINAL

100 CARDINAL

years DATE

5 CARDINAL

years DATE

David PERSON

Greg PERSON

Clark PERSON

OECD ORG

ORG ORG

Graduate PERSON

2007 DATE

87.5 (PERCENT) % PERCENT

UK GPE

GPE ORG

HE GPE

GPE ORG

4 CARDINAL

CARDINAL ORG

10 CARDINAL

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CARDINAL ORG
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Unfinished PERSON

months DATE

3 CARDINAL

CARDINAL ORG

firstly ORDINAL

2020 DATE

2009 DATE

first ORDINAL

2012 DATE

HEPI ORG

ORG ORG

BBC ORG

ODC ONC

ORG ORG

ORG ORG

ComRes ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

ORG ORG

ORG ORG

March DATE

Department (ORG ORG

ORG ORG

BIS ORG

ORG) Department ORG

ORG) DfE ORG

ORG ORG

Department (ORG ORG

ORG ORG

National (ORG) Student (ORG) Survey (ORG ORG

US GPE

GPE ORG

ORG) Authority ORG

ORG ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

three CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CANDINAL OI

3 CARDINAL

years DATE

the 1992 DATE

2006 DATE

2015 DATE

PERCENT ORG

55 DATE

PERCENT ORG

45 DATE

22,000 MONEY

CBI ORG

ORG ORG

Skills (ORG ORG

47 (PERCENT) % PERCENT

Parliament ORG

ORG ORG

David PERSON

ORG ORG

English LANGUAGE

ORG ORG

ORG ORG

TEF ORG

TEF ORG

ORG ORG

TEF ORG

ORG ORG

REF ORG

ORG ORG

ORG) Agency ORG

ORG ORG

2010 DATE

ORG) Department ORG

Skills PRODUCT

CARDINAL ORG

TEF ORG

ORG ORG

UK GPE

GPE ORG

2015 DATE

ORG ORG

2:1 CARDINAL

QUANTITY ORG

QUANTITY ORG

2:1s CARDINAL

CARDINAL ORG

UK GPE

GPE ORG

300 (PERCENT) % PERCENT

1990s DATE

70 DATE

First ORDINAL

2:1 CARDINAL

CARDINAL ORG

CARDINAL ORG

5 CARDINAL

years DATE

PERCENT ORG

47 DATE

mid-1990s DATE

2013 DATE

14 CARDINAL

CARDINAL ORG

PERCENT ORG

second ORDINAL

ORG) Academy ORG

ORG ORG

CARDINAL ORG

CARDINAL ORG

ORG ORG

First ORDINAL

2:1 CARDINAL

CARDINAL ORG

2:2 CARDINAL

CARDINAL ORG

Third ORDINAL

GPA ORG

13 CARDINAL

CARDINAL ORG

2:1 CARDINAL

CARDINAL ORG

ORG ORG

ORG ORG

30.5 CARDINAL

TIME ORG

hours TIME

TIME ORG

14 CARDINAL

TIME ORG

hours TIME

TIME ORG

weekly DATE

Mass (ORG) Communications ORG

ORG ORG

ORG ORG

QUANTITY ORG

QUANTITY ORG

QUANTITY ORG

22 DATE

QUANTITY ORG

hours TIME

QUANTITY ORG

GPA ORG

ORG ORG

Robert PERSON

ORG ORG

UK GPE

GPE ORG

GPA ORG

ORG ORG

Higher PERSON

ORG) ORG

ORG ORG

2012 DATE

2020 DATE

CARDINAL ORG

2009 DATE

Les PERSON

Ebdon PERSON

2 CARDINAL

years DATE

years DATE

November DATE

2016 DATE

GPE ORG

GPE ORG

U.S. GPE

ORG ORG

Senate ORG

New (GPE) Hampshire ORG

GPE ORG

GPE ORG

four CARDINAL

CARDINAL ORG

GPE ORG

CARDINAL ORG

fifteen CARDINAL

CARDINAL ORG

U.S. GPE

GPE ORG

GPE ORG

Massachusetts GPE

GPE ORG

Vermont GPE

GPE ORG

Democrat NORP

GPE ORG

New (GPE) Hampshire ORG

GPE ORG

GPE ORG

ORG ORG

ORG ORG

House ORG

ORG ORG

November DATE

2016 DATE

6,540 CARDINAL

CARDINAL ORG

New (GPE) Hampshire ORG

GPE ORG

n't GPE

State ORG

ORG ORG

GPE ORG

New (GPE) Hampshire ORG

GPE ORG

60 CARDINAL

GPE ORG

6,540 CARDINAL

CARDINAL ORG

GPE ORG

January DATE

2017 DATE

August DATE

30 CARDINAL

2017 DATE

ten CARDINAL

months DATE

CARDINAL ORG

1,014 CARDINAL

CARDINAL ORG

6,540 CARDINAL

CARDINAL ORG

GPE ORG

5,526 CARDINAL

CARDINAL ORG

GPE ORG

5,526 CARDINAL

CARDINAL ORG

CARDINAL ORG

213 CARDINAL

CARDINAL ORG

New (GPE) Hampshire ORG

GPE ORG

5,313 CARDINAL

CARDINAL ORG

GPE ORG

New (GPE) Hampshire ORG

GPE ORG

New (GPE) Hampshire ORG

GPE ORG

State ORG

ORG ORG

5,513 CARDINAL

CARDINAL ORG

two CARDINAL

CARDINAL ORG

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Republican NORP
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NORP PERSON

U.S. GPE

GPE ORG

Kelly PERSON

Ayotte PERSON

Maggie PERSON

Hassan PERSON

Hassan PERSON

1,017 CARDINAL

CARDINAL ORG

5,313 CARDINAL

CARDINAL ORG

59.2 (PERCENT) percent PERCENT

Hassan PERSON

Democrat NORP

GPE ORG

GPE ORG

four CARDINAL

CARDINAL ORG

Hillary PERSON

Clinton PERSON

Clinton PERSON

GPE ORG

CARDINAL ORG

2,732 CARDINAL

CARDINAL ORG

74.8 (PERCENT) percent PERCENT

5,513 CARDINAL

CARDINAL ORG

Clinton PERSON

New (GPE) Hampshire ORG

GPE ORG

three CARDINAL

CARDINAL ORG

four CARDINAL

CARDINAL ORG

Senate ORG

ORG ORG

52 CARDINAL

CARDINAL ORG

48 CARDINAL

Republican NORP

Democrat NORP

Senate ORG

ORG ORG

53 DATE

TIME ORG

47 DATE

TIME ORG

Kris PERSON

State ORG

ORG ORG

Kansas GPE

GPE ORG

Arizona GPE

GPE ORG

10 CARDINAL

CARDINAL ORG

Obama PERSON

2012 DATE

2017 DATE

Trump PERSON

ORG ORG

Kansas GPE

GPE ORG

Miami GPE

GPE ORG

GPE ORG

**Bruce PERSON** 

Mowry PERSON

Miami GPE

GPE ORG

GPE ORG

Biscayne PERSON

Miami GPE

GPE ORG

GPE ORG

400 CARDINAL

2018 DATE

Miami GPE

GPE ORG

GPE ORG

billions ( MONEY

CARDINAL ORG

Harvey (PERSON) Ruvin PERSON

Miami GPE

GPE ORG

GPE ORG

Florida GPE

GPE ORG

Republican NORP

Democrats NORP

James PERSON

Republican NORP

Coral (ORG) Gables (ORG ORG

hundreds CARDINAL

CARDINAL ORG

years DATE

Cason PERSON

South LOC

LOC ORG

Florida GPE

LOC ORG

Vladimir PERSON

Putin PERSON

Europe LOC

LOC ORG

Elisa PERSON

Granato PERSON

first ORDINAL

Oxford ORG

ORG ORG

Europe LOC

LOC ORG

two CARDINAL

Elisa PERSON

TIME ORG

hours TIME

TIME ORG

Elisa PERSON

Granato PERSON

one CARDINAL

two CARDINAL

CARDINAL ORG

Four CARDINAL

CARDINAL ORG

four CARDINAL

CARDINAL ORG

Elisa PERSON

Granato PERSON

BBC ORG

ORG ORG

Elisa PERSON

four CARDINAL

CARDINAL ORG

CARDINAL ORG

99 CARDINAL

CARDINAL ORG

Indonesian NORP

780 CARDINAL

CARDINAL ORG

98.9 CARDINAL

CARDINAL ORG

4.1 CARDINAL

CARDINAL ORG

Covid-19 ORG

ORG ORG

One CARDINAL

CARDINAL ORG

Anglia GPE

Ruskin PERSON

ORG) University ORG

Cambridge GPE

GPE ORG

European NORP

ORG ORG

Vitamin PERSON

May DATE

CARDINAL ORG

100,000 CARDINAL

CARDINAL ORG

Asian NORP

Indonesian NORP

CARDINAL ORG

Prabowo GPE

April DATE

10 CARDINAL

CARDINAL ORG

CARDINAL ORG

five CARDINAL

CARDINAL ORG

British NORP

CARDINAL ORG

six CARDINAL

CARDINAL ORG

CARDINAL ORG

months DATE

Public Health England ORG

VITAMIN ORG

Asian NORP

NHS ORG

ORG ORG

African NORP

African NORP

south Asian NORP

a daily DATE

Vitamin PERSON

BAME PERSON

UK GPE

GPE ORG

William PERSON

University (ORG) of (ORG) Exeter ORG

ORG ORG

MailOnline ORG

ORG ORG

COVID-19 PERSON

ORG ORG

UK GPE

GPE ORG

European NORP

winter DATE

months DATE

African NORP

African NORP

south Asian NORP

Boris PERSON

Johnson PERSON

Tory PERSON

two CARDINAL

**QUANTITY ORG** 

QUANTITY ORG

metre PERSON

**QUANTITY ORG** 

QUANTITY UKG

Oliver PERSON

50 CARDINAL

CARDINAL ORG

Friday DATE

CARDINAL ORG

CARDINAL ORG

1,600 CARDINAL

CARDINAL ORG

months DATE

2.2million CARDINAL

CARDINAL ORG

England GPE

GPE ORG

100 CARDINAL

Vitamin PERSON

3p (PERCENT PERCENT

April DATE

Queen (ORG) Mary (ORG) University ORG

London GPE

GPE ORG

Vitamin PERSON

The Guardian (ORG ORG

TIME ORG

TIME ORG

ORG ORG

Surrey PERSON

ORG ORG

ORG ORG

British NORP

ORG ORG

Dr Lee PERSON

Anglia GPE

Ruskin PERSON

75 DATE

D. NORP

NORP PERSON

NHS ORG

ORG ORG

US GPE

GPE ORG

100 CARDINAL

CARDINAL ORG

Coronavirus PERSON

Vitamin PERSON

Dan PERSON

ORG ORG

November DATE

February DATE

3 CARDINAL

2021 DATE

Dan PERSON

November DATE

CARDINAL ORG

U.S. GPE

GPE ORG

Wisconsin GPE

GPE ORG

ORG ORG

WEC ORG

ORG ORG

234,000 CARDINAL

CARDINAL ORG

Kanye PERSON

ORG ORG

Joe PERSON

Biden PERSON

Milwaukee GPE

GPE ORG

Dane PERSON

GPE ORG

Biden PERSON

WEC ORG

ORG ORG

**Election PERSON** 

November DATE

CARDINAL ORG

zero CARDINAL

CARDINAL ORG

CARDINAL ORG

4,270 CARDINAL

CARDINAL ORG

1,969,274 CARDINAL

CARDINAL ORG

0.2 (PERCENT) % PERCENT

November DATE

2016 DATE

years DATE

2.5 (PERCENT) % PERCENT

first ORDINAL

1,434 CARDINAL

CARDINAL ORG

0.05 (PERCENT) % PERCENT

1.045 CARDINAL

CARDINAL ORG

0.002 (PERCENT) % PERCENT

CARDINAL ORG

42 DATE

CARDINAL ORG

1.8 (PERCENT) % PERCENT

months DATE

Wisconsin GPE

GPE ORG

spring DATE

1,101,474 CARDINAL

CARDINAL ORG

78 (PERCENT) % PERCENT

November DATE

78 (PERCENT) % PERCENT

hundreds CARDINAL

CARDINAL ORG

CARDINAL ORG

thousands CARDINAL

CARDINAL ORG

Wisconsinites NORP

the April DATE

2016 DATE

2.5 (PERCENT) % PERCENT

1.4 (PERCENT) % PERCENT

November DATE

2020 DATE

April DATE

2020 DATE

1.8 (PERCENT) % PERCENT

0.2 (PERCENT) % PERCENT

November DATE

2018 DATE

Milwaukee GPE

GPE ORG

Tony PERSON

Scott PERSON

0.2 (PERCENT) % PERCENT

two CARDINAL

CARDINAL ORG

Republicans NORP

Donald PERSON

ORG ORG

2016 DATE

WEC ORG

ORG ORG

n't GPE

2018 DATE

2020 DATE

CARDINAL ORG

two CARDINAL

CARDINAL ORG

Democrat NORP

Republican NORP

Justine PERSON

23 CARDINAL

June DATE

Britain GPE

GPE ORG

Britain GPE

GPE ORG

Europe LOC

LOC ORG

Britain GPE

GPE ORG

Europe LOC

LOC ORG

Britain GPE

GPE ORG

Europe LOC

LOC ORG

CARDINAL ORG

half CARDINAL

CARDINAL ORG

Europe LOC

LOC ORG

CARDINAL ORG

CARDINAL ORG

four CARDINAL

CARDINAL ORG

CARDINAL ORG

British NORP

Europe LOC

LOC ORG

n't GPE

Britain GPE

GPE ORG

26.5bn MONEY

EU ORG

ORG ORG

years DATE

BMW ORG

ORG ORG

7,000 CARDINAL

CARDINAL ORG

Oxford ORG

GPE ORG

Birmingham GPE

GPE ORG

Swindon ORG

ORG ORG

UK GPE

GPE ORG

Europe LOC

LOC ORG

88 CARDINAL

EU ORG

ORG ORG

ORG) Bank ORG

ORG ORG

Europe LOC

LOC ORG

Europe LOC

LOC ORG

Europe LOC

LOC ORG

Erasmus PERSON

Britain GPE

GPE ORG

Europe LOC

LOC ORG

Britain GPE

GPE ORG

Europe LOC

LOC ORG

EU ORG

ORG ORG

French NORP

German NORP

NORP PERSON

10 CARDINAL

Europe LOC

LOC ORG

Europe LOC

LOC ORG

CARDINAL ORG

London GPE

GPE ORG

2007 DATE

Italy GPE

GPE ORG

Britain GPE

GPE ORG

European NORP

ORG ORG

EU ORG

ORG ORG

CARDINAL ORG

5,000 CARDINAL

CARDINAL ORG

EU ORG

ORG ORG

2010 DATE

European NORP

Britain GPE

GPE ORG

EU ORG

ORG ORG

ORG ORG

Nato ORG

ORG ORG

EU ORG

ORG ORG

Russia GPE

GPE ORG

Ukraine GPE

GPE ORG

LOC ORG

LOC ORG

LOC ORG

Africa LOC

LOC ORG

British NORP

West LOC

Africa LOC

GPE ORG

Europe LOC

LOC ORG

first ORDINAL

EU ORG

ORG ORG

Europe LOC

LOC ORG

Norway GPE

GPE ORG

Switzerland GPE

GPE ORG

Europe LOC

LOC ORG

Russia GPE

GPE ORG

Europe LOC

LOC ORG

EU ORG

ORG ORG

EU ORG

ORG ORG

thousands CARDINAL

CARDINAL ORG

Europe LOC

LOC ORG

Europe LOC

LOC ORG

Britain GPE

GPE ORG

British NORP

Europe LOC

LOC ORG

EU ORG

ORG ORG

790,000 CARDINAL

CARDINAL ORG

UK GPE

GPE ORG

2030 DATE

ORG ORG

Britain GPE

GPE ORG

Europe LOC

LOC ORG

Europe LOC

LOC ORG

Europe LOC

LOC ORG

Britain GPE

GPE ORG

Europe LOC

LOC ORG

Europe LOC

LOC ORG

British NORP

ORG ORG

Britain GPE

Europe LOC

LOC ORG

Biden PERSON

Hayes PERSON

GEORGIA GPE

GPE ORG

Donald PERSON

2015 DATE

Muslims NORP

GPE ORG

Europe LOC

LOC ORG

ISIS ORG

ORG ORG

weekly DATE

hundreds CARDINAL

CARDINAL ORG

85 DATE

CARDINAL ORG

Paris GPE

GPE ORG

German NORP

NORP PERSON

Christian NORP

Islam ORG

Islam ORG

ORG ORG

GPE ORG

Georgia GPE

GPE ORG

Muslim NORP

Jason PERSON

Wednesday DATE

Spencer PERSON

Islamic NORP

Spencer PERSON

Islamic NORP

GPE ORG

GPE ORG

Georgia GPE

GPE ORG

Spencer PERSON

WSB ORG

ORG ORG

Atlanta GPE

GPE ORG

LAW ORG

Islam ORG

ORG ORG

Christians NORP

Bible WORK OF ART

Quran PERSON

GPE ORG

**Ouran PERSON** 

GPE ORG

Allah (GPE ORG

Messenger PERSON

Islam NORP

Jews NORP

Christians NORP

Jizya PERSON

LOC ORG

Allah (GPE ORG

Messenger PERSON

Islam NORP

Jews NORP

Christians NORP

Jizya PERSON

LOC ORG

Quran PERSON

GPE ORG

Allah (ORG ORG

Christian NORP

ORG ORG

GPE ORG

Georgia GPE

GPE ORG

Islam ORG

France GPE

GPE ORG

Belgium GPE

GPE ORG

Italy GPE

GPE ORG

Spain GPE

GPE ORG

Switzerland GPE

GPE ORG

American NORP

Trump PERSON

American NORP

GEORGIA PERSON

GPE ORG

Control (ORG ORG

ORG ORG

Bill PERSON

41 CARDINAL

CARDINAL ORG

ORG ORG

House ORG

ORG ORG

Nigerians NORP

ORG ORG

House ORG

ORG ORG

ORG ORG

Bill PERSON

ORG ORG

second ORDINAL

82 CARDINAL

```
CARDINAL ORG
House ORG
ORG ORG
Femi Gbajabiamila ORG
ORG ORG
NCDC ORG
ORG ORG
NCDC ORG
ORG ORG
Nigeria GPE
GPE ORG
Tuesday DATE
NCDC ORG
ORG ORG
NCDC ORG
ORG ORG
3 CARDINAL
CARDINAL ORG
Nigeria PERSON
GPE ORG
Centre for Democracy and Development , Civil (ORG) ORG
Advocacy PERSON
Say No Campaign ORG
Amnesty ORG
YIAGA ORG
GPE ORG
Africa LOC
GPE ORG
Agenda PERSON
ORG ORG
Global (ORG) Rights (ORG ORG
African NORP
Alert PERSON
Women (ORG) ORG
ORG ORG
ORG) Community ORG
ORG) Centre ORG
ORG) Centre ORG
Advocacy PERSON
Concerned (ORG) Nigerians (ORG ORG
ORG ORG
Education ORG
ORG ORG
Vaccine PERSON
Dinidari PERSON
ORG ORG
Albino PERSON
ORG ORG
ORG ORG
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ORG ORG

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Nigeria GPE
GPE ORG
Dorothy PERSON
ORG) ORG
ORG) Foundation ORG
ORG ORG
Raising (ORG) WORK OF ART
ORG ORG
Youths PERSON
ORG ORG
Adopt A Goal For Development Initiative , Coalition in Defence of
Nigerian Democracy and Constitution WORK_OF_ART
GPE ORG
House ORG
ORG ORG
Molluma PERSON
Medico PERSON
Albino PERSON
ORG ORG
ORG ORG
ORG ORG
Nigeria GPE
GPE ORG
GPE ORG
Africa LOC
LOC ORG
Youths PERSON
Motion (ORG ORG
ORG ORG
ORG ORG
Silverchip Fox PERSON
Tuesday DATE
Gbajabiamila PERSON
Nigerians NORP
Nigerian NORP
House ORG
ORG ORG
Tesco PERSON
ORG ORG
119 CARDINAL
CARDINAL ORG
Robin PERSON
Williams PERSON
ORG ORG
16 CARDINAL
CARDINAL ORG
CARDINAL ORG
24 CARDINAL
```

12 CARDINAL months DATE

Britain GPE

GPE ORG

Europe LOC

LOC ORG

Young PERSON

Generation A PRODUCT

LOC ORG

LOC ORG

QUANTITY ORG

2 CARDINAL

QUANTITY ORG

QUANTITY ORG

CNN ORG

ORG ORG

Monday DATE

ORG) Sciences ORG

ORG ORG

University (ORG ORG

ORG ORG

Colorado GPE

ORG ORG

Steve PERSON

Nerem PERSON

1993 DATE

1992 DATE

2014 DATE

QUANTITY ORG

QUANTITY ORG

2.8 CARDINAL

QUANTITY ORG

QUANTITY ORG

25 CARDINAL

years DATE

CARDINAL ORG

3 CARDINAL

CARDINAL ORG

0.1 CARDINAL

**OUANTITY ORG** 

QUANTITY ORG

Leon PERSON

Young PERSON

GPE ORG

three CARDINAL

months DATE

75 (PERCENT) percent PERCENT

12,000 MONEY

49,943 MONEY

2011 DATE

12 CARDINAL

34 CARDINAL

122 CARDINAL

Young (PERSON WORK\_OF\_ART

2012 DATE

Young PERSON

seven CARDINAL

CARDINAL ORG

seven CARDINAL

Wisconsin GPE

GPE ORG

10 CARDINAL

CARDINAL ORG

Milwaukee GPE

ORG) Board ORG

ORG ORG

Republicans NORP

Legislature ORG

Alberta PERSON

Joe PERSON

15,000 MONEY

51,000 MONEY

**\$ (MONEY MONEY** 

6.5 CARDINAL

Legislature ORG

ORG ORG

Milwaukee GPE

ORG) Board ORG

ORG ORG

Madison PERSON

Joyce PERSON

ORG ORG

16 CARDINAL

years DATE

Mike PERSON

232,000 MONEY

Joyce PERSON

ORG ORG

annual DATE

430,000 MONEY

ORG ORG

Capitol FAC

FAC ORG

ORG ORG

McCabe (ORG) 's ORG

McCabe (ORG ORG

Capitol ORG

Michigan GPE

GPE ORG

Michigan GPE

GPE ORG

Alcona GPE

GPE ORG

the 73rd DATE

1998 DATE

ORG ORG

Alcona GPE

GPE ORG

Alpena ORG

GPE ORG

Montmorency PERSON

Oscoda ORG

TB ORG

ORG ORG

Michigan GPE

ORG) Department ORG

ORG ORG

Michigan GPE

GPE ORG

Fox PERSON

ORG ORG

GPE ORG

Ala. GPE

GPE ORG

State ORG

Veterinarian Nancy PERSON

Barr PERSON

ORG ORG

5 CARDINAL

Muslim NORP

Yemen GPE

GPE ORG

Hunger (ORG) ORG

COVID-19 PERSON

ORG) & (ORG) Poverty ORG

ORG ORG

2020 DATE

two CARDINAL

years DATE

2018 DATE

2020 DATE

85,000 CARDINAL

CARDINAL ORG

Yemen GPE

GPE ORG

2018 DATE

CARDINAL ORG

UK GPE

GPE ORG

5 CARDINAL

CARDINAL ORG

CARDINAL ORG

```
Yemen GPE
```

GPE ORG

2018 DATE

2020 DATE

UK GPE

GPE ORG

CARDINAL ORG

CARDINAL ORG

5 CARDINAL

CARDINAL ORG

CARDINAL ORG

Yemen GPE

GPE ORG

2020 DATE

COVID-19 PERSON

Muslims NORP

Madina GPE

GPE ORG

COVID-19 PERSON

ORG ORG

Helle PERSON

Schmidt PERSON

Save (ORG) the (ORG) Children (ORG) International ORG

Yemen GPE

GPE ORG

ORG) & (ORG ORG

ORG ORG

COVID-19 PERSON

ORG ORG

2018 DATE

four CARDINAL

CARDINAL ORG

CARDINAL ORG

Save (ORG) ORG

ORG ORG

Wednesday DATE

CARDINAL ORG

CARDINAL ORG

almost 2 CARDINAL

years DATE

Yemen GPE

GPE ORG

CARDINAL ORG

Yemen GPE

GPE ORG

CARDINAL ORG

Yemen GPE

GPE ORG

Schmidt PERSON

Telegraph ORG

FAC ORG

UK GPE

GPE ORG

2018 DATE

5.2 CARDINAL

CARDINAL ORG

CARDINAL ORG

Yemen GPE

GPE ORG

Britain GPE

GPE ORG

NGO ORG

ORG ORG

2018 DATE

CARDINAL ORG

CARDINAL ORG

10 CARDINAL

CARDINAL ORG

Yemen GPE

GPE ORG

Yemen GPE

GPE ORG

CARDINAL ORG

CARDINAL ORG

85,000 CARDINAL

CARDINAL ORG

Guardian ORG

CARDINAL ORG

CARDINAL ORG

five CARDINAL

CARDINAL ORG

CARDINAL ORG

Yemen GPE

GPE ORG

Yemen GPE

GPE ORG

Yemen GPE

GPE ORG

Muslim NORP

Arab NORP

Saudi NORP

GPE ORG

Muslim NORP

Ummah PERSON

Muhammad PERSON

TIME ORG

20 CARDINAL

TIME ORG

hours TIME

TIME ORG

Friday DATE

Breonna PERSON

Taylor PERSON

Taylor PERSON

today DATE

Taylor PERSON

Taylor PERSON

a months DATE

months DATE

Taylor PERSON

ORG ORG

Taylor PERSON

Taylor PERSON

Officer Brett PERSON

Hankison PERSON

Taylor PERSON

Kenneth PERSON

CARDINAL ORG

Taylor PERSON

Taylor PERSON

TIME ORG

TIME ORG

TIME ORG

TIME ORG

TIME ORG

TIME ORG

15 CARDINAL

TIME ORG

seconds TIME

TIME ORG

Louisville GPE

GPE ORG

Kentucky GPE

GPE ORG

Shawn PERSON

Taylor PERSON

third ORDINAL

45 DATE

TIME ORG

seconds TIME

TIME ORG

TIME ORG

TIME ORG

Walker (ORG ORG

CARDINAL ORG

Taylor PERSON

4 CARDINAL

CARDINAL ORG

n't GPE

Breonna PERSON

Taylor PERSON

Taylor PERSON

Louisville GPE

GPE ORG

March DATE

13 CARDINAL

26 CARDINAL

five CARDINAL

CARDINAL ORG

Cameron PERSON

ORG ORG

two CARDINAL

CARDINAL ORG

Taylor PERSON

Taylor PERSON

Officer Brett PERSON

Hankison PERSON

three CARDINAL

CARDINAL ORG

American NORP

Taylor PERSON

Jason PERSON

Taylor PERSON

Breonna PERSON

Taylor PERSON

SHE ORG

ORG ORG

BILL PERSON

ORG ORG

ORG ORG

ORG ORG

WILL LIKELY USE PERSON

EXPERIMENTAL ORG

first ORDINAL

COVID-19 PERSON

ORG ORG

n't GPE

COVID-19 PERSON

ORG ORG

RNA ORG

ORG ORG

COVID-19 ORG

ORG ORG

n't ORG

Moderna PERSON

GPE ORG

RNA ORG

ORG ORG

thirty CARDINAL

years DATE

5 CARDINAL

CARDINAL ORG

two CARDINAL

CARDINAL ORG

RNA ORG

ORG ORG

Coronavirus LOC

LOC ORG

U.S. GPE

GPE ORG

Chicago GPE

GPE ORG

hundreds CARDINAL

Americans NORP

Bill PERSON

WHO ORG

ORG ORG

years DATE

Gates ORG

ORG ORG

7 (CARDINAL WORK\_OF\_ART

CARDINAL ORG

RNA ORG

ORG ORG

Gates ORG

ORG ORG

RNA ORG

ORG ORG

two CARDINAL

CARDINAL ORG

COVID-19 ORG

ORG ORG

QUANTITY ORG

QUANTITY ORG

two CARDINAL

CARDINAL ORG

RNA ORG

ORG ORG

CARDINAL ORG

RNA ORG

ORG ORG

CEPI ORG

ORG ORG

RNA ORG

ORG ORG

CARDINAL ORG

COVID ORG

first ORDINAL

RNA ORG

ORG ORG

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Moderna PERSON
GPE ORG
RNA ORG
ORG ORG
RNA ORG
ORG ORG
RNA ORG
ORG ORG
n't GPE
n't GPE
RNA ORG
ORG ORG
COVID ORG
ORG ORG
first ORDINAL
RNA ORG
ORG ORG
first ORDINAL
Bill PERSON
ORG ORG
CARDINAL ORG
Donald PERSON
Cabinet ORG
Republican NORP
Trump ORG
ORG ORG
Hillary PERSON
Clinton PERSON
Gingrich PERSON
two CARDINAL
```

years DATE Trump ORG ORG ORG four CARDINAL years DATE

the Electoral (ORG) College ORG

2012 DATE
Mitt PERSON
Romney PERSON
Barack PERSON
Obama PERSON
Trump PERSON
ORG ORG
April DATE

Ben PERSON
Carson PERSON
Gingrich PERSON
Trumpsters ORG

ORG ORG Trump ORG

2014 DATE

Gingrich PERSON

National (ORG) Popular (ORG) Vote (ORG ORG

11 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

12 CARDINAL

CARDINAL ORG

Americans NORP

two CARDINAL

years DATE

Gingrich PERSON

Gingrich PERSON

Americans NORP

Gingrich PERSON

314 CARDINAL

CARDINAL ORG

CARDINAL ORG

Earth LOC

LOC ORG

Gingrich PERSON

America GPE

GPE ORG

Clinton PERSON

Trump PERSON

ORG ORG

60 CARDINAL

CARDINAL ORG

CARDINAL ORG

59.8 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

fifth ORDINAL

CARDINAL ORG

Americans NORP

n't ORG

Gingrich PERSON

Trump ORG

ORG ORG

CARDINAL ORG

FAC ORG

FAC ORG

House ORG

FAC ORG

Trump ORG

2012 DATE

GPE ORG

Newt PERSON

RT (ORG) - (ORG

ORG ORG

Torsten PERSON

Konstantin PERSON

Demeter PERSON

PCR ORG

ORG ORG

Unfounded "Test PERSON

COVID-19 ORG

ORG ORG

March DATE

16 CARDINAL

2020 DATE

WHO ORG

ORG ORG

Dr Tedros PERSON

Reuters ORG

ORG ORG

BBC ORG

ORG ORG

3 CARDINAL

CARDINAL ORG

May DATE

German NORP

NORP PERSON

PCR ORG

ORG ORG

Walter PERSON

Lippmann PERSON

two CARDINAL

CARDINAL ORG

Pulitzer PERSON

Kary PERSON

1993 DATE

74 DATE

PCR ORG

ORG ORG

PCR ORG

ORG ORG

millions CARDINAL

PCR ORG

ORG ORG

Gina PERSON

2007 DATE

ORG ORG

Faith PERSON

PCR ORG

ORG ORG

COVID-19 ORG

Australian NORP

Sanjaya PERSON

ABC ORG

ORG ORG

decades DATE

COVID-19 PERSON

n't ORG

Jessica PERSON

Watson PERSON

Bristol ORG

ORG) University ORG

ORG ORG

British NORP

ORG ORG

COVID-19 PERSON

ORG ORG

COVID-19 PERSON

Watson PERSON

PCR ORG

ORG ORG

PCR ORG

ORG ORG

PCR ORG

ORG ORG

COVID-19 PERSON

ORG ORG

Thomas PERSON

ORG) Department ORG

ORG ORG

German NORP

COVID-19 PERSON

ORG ORG

COVID-19 PERSON

Watson PERSON

Watson PERSON

Watson PERSON

COVID-19 PERSON

COVID-19 PERSON

ORG ORG

n't PERSON

June DATE

2nd DATE

RNA ORG

ORG ORG

first ORDINAL

RNA ORG

ORG ORG

```
PCR ORG
```

Medical Virology ORG

1986 DATE

9 CARDINAL

CARDINAL ORG

Luc PERSON

Nobel WORK OF ART

Marie PERSON

100 CARDINAL

CARDINAL ORG

1898 DATE

RNA ORG

ORG ORG

PCR ORG

ORG ORG

PCR ORG

ORG ORG

RNA ORG

ORG ORG

DALA ODG

RNA ORG

ORG ORG

NB ORG

ORG ORG

Malik PERSON

Nature (WORK\_OF\_ART) Medicine (WORK\_OF\_ART WORK\_OF\_ART

March DATE

2020 DATE

Malik PERSON

12 CARDINAL

2020 DATE

2 CARDINAL

CARDINAL ORG

Myung PERSON

Han NORP

ORG ORG

Patient GPE

Korea GPE

GPE ORG

COVID-19 ORG

ORG ORG

Osong PERSON

February DATE

2020 DATE

Myung PERSON

Han NORP

6 CARDINAL

2020 DATE

3 CARDINAL

CARDINAL ORG

Wan PERSON

ORG ORG

Korea GPE

GPE ORG

Journal ORG

February DATE

24 CARDINAL

2020 DATE

March DATE

19 CARDINAL

2020 DATE

4 CARDINAL

CARDINAL ORG

Coronavirus PERSON

Pneumonia GPE

China GPE

GPE ORG

2019 DATE

February DATE

20 CARDINAL

2020 DATE

Wenjie PERSON

Tan PERSON

March DATE

18 CARDINAL

2020 DATE

four CARDINAL

CARDINAL ORG

2020 DATE

Dr Charles PERSON

2001 DATE

Calisher PERSON

3 CARDINAL

Dr PERSON

CARDINAL ORG

4 CARDINAL

RNA ORG

ORG ORG

PCR ORG

ORG ORG

RNA ORG

ORG ORG

COLUTE 10

COVID-19 ORG

ORG ORG

CARDINAL ORG

four CARDINAL

CARDINAL ORG

Koch PERSON

Amory ORG

Devereux ORG

Rosemary PERSON

OffGuardian NORP

CARDINAL ORG

Nature WORK OF ART

GPE ORG

May DATE

third ORDINAL

German NORP

NORP PERSON

Alexander PERSON

Hartmut PERSON

Hengel PERSON

Ralf PERSON

German NORP

ORG ORG

Thomas PERSON

Ulrich PERSON

Charité Berlin PERSON

Bornkamm PERSON

RNA ORG

ORG ORG

Nobel WORK OF ART

Barbara PERSON

1983 DATE

Christian (ORG) ORG

Germany GPE

GPE ORG

COVID-19 ORG

ORG ORG

German NORP

NORP PERSON

PCR ORG

ORG ORG

first ORDINAL

WHO ORG

ORG ORG

n't PERSON

June DATE

18 CARDINAL

2020 DATE

months DATE

Berlin GPE

GPE ORG

Charité NORP

n't GPE

Corman (NORP ORG

RNA ORG

ORG ORG

Roche PERSON

Penzberg PERSON

GPE ORG

Germany GPE

GPE ORG

RNA ORG

ORG ORG

mini kit QIAGEN PERSON

ORG ORG

Hilden GPE

GPE ORG

Germany GPE

GPE ORG

RNA ORG

ORG ORG

January DATE

23 CARDINAL

2020 DATE

n't GPE

two CARDINAL

CARDINAL ORG

PCR ORG

ORG ORG

February DATE

China GPE

GPE ORG

Guangdong GPE

GPE ORG

GPE ORG

COVID-19 ORG

ORG ORG

ORG ORG

29 CARDINAL

CARDINAL ORG

610 CARDINAL

CARDINAL ORG

Wuhan GPE

GPE ORG

3 CARDINAL

CARDINAL ORG

6 CARDINAL

CARDINAL ORG

A third CARDINAL

Singapore GPE

GPE ORG

18 CARDINAL

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

five CARDINAL

CARDINAL ORG

CARDINAL ORG

Wang PERSON

Chen PERSON

Chinese NORP

ORG ORG

February DATE

PCR ORG

ORG ORG

CARDINAL ORG

30 CARDINAL

CARDINAL ORG

CARDINAL ORG

50 CARDINAL

CARDINAL ORG

Sin PERSON

Lee PERSON

ORG ORG

WHO ORG

ORG ORG

Anthony PERSON

March DATE

22 DATE

2020 DATE

RNA ORG

ORG ORG

PCR ORG

ORG ORG

PPV ORG

ORG ORG

PPV ORG

ORG ORG

two CARDINAL

CARDINAL ORG

95 (PERCENT) % PERCENT

5 CARDINAL

CARDINAL ORG

100 CARDINAL

CARDINAL ORG

PPV ORG

ORG ORG

June DATE

12 CARDINAL

2020 DATE

Deutsches PERSON

PPV ORG

ORG ORG

three CARDINAL

CARDINAL ORG

second ORDINAL

Jessica PERSON

Watson PERSON

95 (PERCENT) % PERCENT

Deutsches PERSON

PPV ORG

ORG ORG

CARDINAL ORG

three CARDINAL

CARDINAL ORG

3 (PERCENT) % PERCENT

PPV ORG

ORG ORG

PERCENT ORG

30 (PERCENT) percent PERCENT

70 (PERCENT) percent PERCENT

20 (PERCENT) percent PERCENT

PERCENT GPE

PPV ORG

ORG ORG

78 (PERCENT) percent PERCENT

22 (PERCENT) percent PERCENT

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

2 CARDINAL

CARDINAL ORG

CARDINAL ORG

CDC ORG

ORG ORG

FDA ORG

ORG ORG

CDC ORG

ORG ORG

2019 DATE

2019 DATE

March DATE

30 CARDINAL

2020 DATE

RNA ORG

ORG ORG

2019 DATE

nCoV ORG

FDA ORG

ORG ORG

PCR ORG

ORG ORG

Altona PERSON

ORG ORG

LightMix ORG

TIB ORG

Molbiol PERSON

ORG ORG

Roche ORG

ORG ORG

PCR ORG

ORG ORG

Roche ORG

ORG ORG

ORG ORG

n't GPE

millions CARDINAL

CARDINAL ORG

CARDINAL ORG

millions CARDINAL

CARDINAL ORG

CDC ORG

ORG ORG

WHO ORG

ORG ORG

FDA ORG

ORG ORG

ORG ORG

Jon PERSON

RNA ORG

ORG ORG

PCR ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

CARDINAL ORG

PCR ORG

ORG ORG

29 CARDINAL

CARDINAL ORG

86 DATE

199 CARDINAL

CARDINAL ORG

272 CARDINAL

CARDINAL ORG

293 CARDINAL

CARDINAL ORG

un ORG

ORG ORG

Viviane PRODUCT

RT (ORG ORG

ORG ORG

Asian NORP

Asian NORP

```
April DATE
WHO ORG
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Asian NORP

Roche ORG

ORG ORG

TIB ORG

Molbiol PERSON

ORG ORG

Cq PERSON

PCR ORG

ORG ORG

35 DATE

45 CARDINAL

CARDINAL ORG

Cq PERSON

CARDINAL ORG

CARDINAL ORG

40 DATE

CARDINAL ORG

ORG ORG

MIQE PERSON

ORG ORG

ORG ORG

PCR PERSON

ORG ORG

NORP PERSON

Kary PERSON

CARDINAL ORG

CARDINAL ORG

40 DATE

CARDINAL ORG

ORG ORG

ORG ORG

Stephen GPE

PCR PERSON

ORG ORG

A (WORK\_OF\_ART) - (WORK\_OF\_ART) Z WORK\_OF\_ART

WORK\_OF\_ART ORG

NORP PERSON

Bustin PRODUCT

Cq PERSON

20s DATE

30s DATE

Cq PERSON

35 DATE

GPE ORG

PCR ORG

ORG ORG

RNA ORG

```
ORG ORG
RNA ORG
ORG ORG
Reverse PERSON
Jessica PERSON
Toronto GPE
GPE ORG
two CARDINAL
CARDINAL ORG
2019 DATE
Stephen PERSON
PCR ORG
ORG ORG
RNA ORG
ORG ORG
RNA ORG
ORG ORG
10 CARDINAL
CARDINAL ORG
PCR ORG
ORG ORG
COVID-19 PERSON
NB ORG
ORG ORG
WHO ORG
ORG ORG
ORG ORG
British NORP
ORG ORG
2010 DATE
WHO ORG
ORG ORG
CDC ORG
ORG ORG
these RT (ORG) - ( ORG
ORG ORG
Fatal (WORK_OF_ART) Therapie WORK_OF_ART
ORG ORG
COVID-19 ORG
ORG ORG
1 CARDINAL
ORG) Sensitivity ORG
ORG ORG
2 CARDINAL
CARDINAL ORG
Thomas PERSON
March DATE
6 CARDINAL
2020 DATE
3 CARDINAL
```

```
CARDINAL ORG
Martin Enserink PERSON
Science (ORG) ORG
July DATE
6 CARDINAL
2001 DATE
24 CARDINAL
CARDINAL ORG
4 CARDINAL
CARDINAL ORG
Charles PERSON
May DATE
10 CARDINAL
2020 DATE
5 CARDINAL
CARDINAL ORG
Torsten Engelbrecht PERSON
Hamburg GPE
GPE ORG
Germany GPE
GPE ORG
2006 DATE
ORG ORG
Klaus PERSON
Kohnlein PERSON
2009 DATE
German NORP
NORP PERSON
Rubikon PERSON
Süddeutsche ORG
ORG) Zeitung ORG
Financial (ORG) ORG
ORG ORG
Konstantin PERSON
Demeter PERSON
Torsten PERSON
ORG ORG
Rubikon PERSON
Swiss GPE
Italian NORP
```

Application de la fonction précédente à la colonne "title" avec la méthode apply

```
# Définir une fonction pour ajouter le nom de l'entité à chaque entité
nommée dans la colonne de texte
def add_entity_name(text):
    doc = nlp(text)
    # Créer une liste pour stocker les nouveaux tokens
    nouveaux tokens = []
```

```
# Parcourir les tokens et ajouter des informations selon le type
d'entité
    for token in doc:
        if token.ent type :
             # Ajouter le nom de l'entité (type d'entité) à côté de
l'entité
             nouveaux tokens.append(f"{token.text}
({token.ent type })")
        else:
             nouveaux tokens.append(token.text)
    # Reconstruire le texte avec les informations ajoutées
    nouveau texte = " ".join(nouveaux tokens)
    return nouveau texte
 #Application de la fonction à la colonne "titre text"
#dftrain["entities"] = dftrain["text_title"].apply(extract_entities)
#print(dftrain["entities"])
dftrain['text title'].fillna('', inplace=True)
dftrain['text title'] = dftrain['text title'].apply(add entity name)
On scinde les données de la colonne "title" en jeu d'apprentissage et jeu de test
X=dftrain['text title']
X_train,X_test,y_train,y_test=train_test split(X,y,test size =
0.2, random state=8)
print("X train is \n", X train.shape)
print("y_train is \n",y_train.shape)
print("X_test is \n",X_test.shape)
print("y test is \n",y test.shape)
X train is
 (675,)
y train is
 (675,)
X test is
 (169,)
y test is
 (169,)
```

Ici, c'est une étape importante, on va tester différents classifieurs, pour chacun des classifieurs, on va appliquer le prétraitement + Vectorisation TfIdf, et on applique une cross\_val\_score avec un Kfold de 10 fois, par la suite on stocke dans une liste all\_results la moyenne des accuracy + l'écart type et on la trie par ordre décroissant de moyenne d'accuracy et d'écart type. on remarque que les 2 meilleurs sont SVM et RF qu'on va séléctionner pour leur appliquer le GridSearch sur les paramètres des prétraitements + leurs hyperparamètres pour pouvoir choisir le meilleur.

```
score = 'accuracy'
seed = 7
allresults = []
results = []
names = []
# Liste des modèles à tester
models = [
    ('MultinomialNB', MultinomialNB()),
    ('LogisticRegression', LogisticRegression(random state=42))
]
models.append(('KNN', KNeighborsClassifier()))
models.append(('CART', DecisionTreeClassifier(random_state=42)))
models.append(('RF', RandomForestClassifier(random state=42)))
models.append(('SVM', SVC(random state=42)))
# Création d'un pipeline pour chaque modèle
pipelines = []
for name, model in models:
    pipeline = Pipeline([
        ('normalize', TextNormalizer()),
        ('tfidf', TfidfVectorizer()),
        (name, model)
    1)
    pipelines.append((name,pipeline))
    #pipeline.fit(X train,y train)
all_results=[]
scores=[]
for p in pipelines:
    print(p[1])
    # cross validation en 10 fois
    kfold = KFold(n splits=10, random state=seed, shuffle=True)
    print ("Evaluation de ",p)
    start time = time.time()
    # application de la classification
    cv_results = cross_val_score(p[1],X_train,y_train, cv=kfold,
scoring=score)
    #print("Pour le classifieur",p[0],"on a un score
de",cv_results.mean(),"et un écart type de",cv results.std())
    scores.append(cv results)
    all_results.append((p[0],cv_results.mean(),cv_results.std()))
    end time = time.time()
```

```
print("all resultats", all results)
all results = sorted(all results, key=lambda x: (-x[1], -x[2]))
print("all resultats", all results)
    # affichage des résultats
#print ('\nLe meilleur resultat : ',max(results))
X train (675,)
y train (675,)
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                 ('MultinomialNB', MultinomialNB())])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                 ('LogisticRegression',
LogisticRegression(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                 ('KNN', KNeighborsClassifier())])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                 ('CART', DecisionTreeClassifier(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                 ('RF', RandomForestClassifier(random state=42))])
Pipeline(steps=[('normalize', TextNormalizer()), ('tfidf',
TfidfVectorizer()),
                 ('SVM', SVC(random state=42))])
all resultats [('MultinomialNB', 0.7879719051799824,
0.08755580534737224), ('LogisticRegression', 0.7852941176470589,
0.04977379242498178), ('KNN', 0.6605575065847236,
0.059506638746038246), ('CART', 0.8237269534679544,
0.06278583197362432), ('RF', 0.902194907813872, 0.033821414200847146), ('SVM', 0.8593942054433713, 0.042467036450607024)]
all resultats [('RF', 0.902194907813872, 0.033821414200847146),
('SVM', 0.8593942054433713, 0.042467036450607024), ('CART',
0.8237269534679544, 0.06278583197362432), ('MultinomialNB',
0.7879719051799824, 0.08755580534737224), ('LogisticRegression',
0.7852941176470589, 0.04977379242498178), ('KNN', 0.6605575065847236,
0.059506638746038246)]
```

On a un pipeline pour chaque prétraitement différent, on essaye pas mal (miniscule, lemmatisation, miniscule + lemmatisation..) et on stocke le fit\_transorm de nos X\_train, X\_test sur les pipelines dans des listes qui vont contenir tous les fit\_transform des pipelines pour chaque classifieur, par la suite on parcourt ces listes là, on itère dessus, et chaque élement de la liste (train) va passer par le GridSearch et puis on predict sur son corresapondant dans liste (test).

```
# le plus simple est de faire un test sur differents pipelines.
# pipeline de l'utilisation de CountVectorizer sur le texte avec
differents pre-traitements
CV_brut = Pipeline([('cleaner', TextNormalizer()),
                    ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowcase = Pipeline([('cleaner',
TextNormalizer(removestopwords=False,lowercase=True,
getstemmer=False, removedigit=False)),
                    ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowStop = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=False,removedigit=False)),
                    ('count vectorizer',
CountVectorizer(lowercase=False))])
CV lowStopstem = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=True, removedigit=False)),
                    ('count vectorizer',
CountVectorizer(lowercase=False))])
# pipeline de l'utilisation de TfidfVectorizer avec differents pre-
traitements
TFIDF brut = Pipeline ([('cleaner', TextNormalizer()),
                    ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowcase = Pipeline([('cleaner',
TextNormalizer(removestopwords=False,lowercase=True,
getstemmer=False, removedigit=False)),
                    ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowStop = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
getstemmer=False, removedigit=False)),
                    ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
TFIDF lowStopstem = Pipeline([('cleaner',
TextNormalizer(removestopwords=True,lowercase=True,
```

```
getstemmer=True,removedigit=False)),
                      ('tfidf vectorizer',
TfidfVectorizer(lowercase=False))])
# Liste de tous les modeles à tester
all models = [
    ("CV brut", CV brut),
    ("CV_lowcase", CV_lowcase),
("CV_lowStop", CV_lowStop),
    ("CV_lowStopstem", CV_lowStopstem),
    ("TFIDF_lowcase", TFIDF_lowcase),
("TFIDF_lowStop", TFIDF_lowStop),
    ("TFIDF_lowStopstem", TFIDF_lowStopstem),
    ("TFIDF brut", TFIDF brut)
1
X train text SVC = []
X \text{ test text SVC} = []
X train text RandomForestClassifier = []
X test text RandomForestClassifier = []
for name, pipeline in all models :
    X_train_text_SVC.append(pipeline.fit_transform(X_train).toarray())
    X test text SVC.append(pipeline.transform(X test).toarray())
X train text RandomForestClassifier.append(pipeline.fit transform(X tr
ain).toarray())
X test text RandomForestClassifier.append(pipeline.transform(X test).t
oarray())
models = {
    'SVC': SVC(random state=42),
    'RandomForestClassifier': RandomForestClassifier(random state=42)
}
params = \{'SVC': [\{'C': [0.001, 0.01, 0.1, 1,2,5,7,10]\},
              {\text{gamma': [0.001, 0.01, 0.1,0.2,0.3,0.5,0.7,1]}},
              {'kernel': ['linear', 'rbf']}],
    'RandomForestClassifier': [{'n_estimators': [10, 50, 100, 200,
3001},
                                {'max features': ['auto', 'sqrt',
```

```
'log2']}]
for model name, model in models.items():
    score='accuracy'
    X_train_text = eval('X_train_text_' + model_name)
    X test text = eval('X test text ' + model name)
    for i in range (len(X train text)):
      grid search = GridSearchCV(model, params[model name], n jobs=-1,
verbose=1,scoring=score)
      print("grid search fait")
      print("X_train", X_train_text[i].shape)
print("y_train", y_train.shape)
      grid search.fit(X train text[i],y train)
      print ('meilleur score %0.3f'%(grid search.best score ),'\n')
      print ('meilleur estimateur',grid_search.best_estimator_,'\n')
      y_pred = grid_search.predict(X_test_text[i])
      MyshowAllScores(y test,y pred)
      print("Ensemble des meilleurs paramètres :")
      best parameters = grid search.best estimator_.get_params()
      for param dict in params[model name]:
        for param name, param value in param dict.items():
            print("\t^s: \t^m \%) (param name,
best parameters[param name]))
y train (675,)
y_test (169,)
X test (169,)
grid search fait
X train (675, 25358)
v train (675,)
Fitting 5 folds for each of 18 candidates, totalling 90 fits
meilleur score 0.839
meilleur estimateur SVC(kernel='linear', random state=42)
Accuracy: 0.852
Classification Report
              precision
                            recall f1-score
                                               support
                0.91429
                           0.77108
                                     0.83660
       FALSE
                                                     83
        TRUE
                0.80808
                           0.93023
                                     0.86486
                                                     86
                                     0.85207
                                                    169
    accuracy
                0.86118
                           0.85066
                                     0.85073
   macro avg
                                                    169
weighted avg
                0.86024
                           0.85207
                                     0.85098
                                                    169
Ensemble des meilleurs paramètres :
     C: 1.0
```

gamma: 'scale'
kernel: 'linear'

grid search fait

X train (675, 21325)

y\_train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits

meilleur score 0.837

meilleur estimateur SVC(gamma=0.01, random\_state=42)

Accuracy: 0.905

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.83838 1.00000	1.00000 0.81395	0.91209 0.89744	83 86
accuracy macro avg weighted avg	0.91919 0.92063	0.90698 0.90533	0.90533 0.90476 0.90463	169 169 169

Ensemble des meilleurs paramètres :

C: 1.0

gamma: 0.01
kernel: 'rbf'

grid search fait

X\_train (675, 21197)

y train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.847

meilleur estimateur SVC(kernel='linear', random\_state=42)

Accuracy: 0.864

Classification Report

support	f1-score	recall	precision	0 (0331) 100 (10
83 86	0.85161 0.87432	0.79518 0.93023	0.91667 0.82474	FALSE TRUE
169 169 169	0.86391 0.86296 0.86317	0.86271 0.86391	0.87070 0.86989	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

C: 1.0

gamma: 'scale'
 kernel: 'linear'
grid search fait

```
X train (675, 14382)
y train (675,)
```

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.840

meilleur estimateur SVC(kernel='linear', random state=42)

Accuracy: 0.870

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.91781 0.83333	0.80723 0.93023	0.85897 0.87912	83 86
accuracy macro avg weighted avg	0.87557 0.87482	0.86873 0.86982	0.86982 0.86905 0.86923	169 169 169

Ensemble des meilleurs paramètres :

C: 1.0

gamma: 'scale' kernel: 'linear'

grid search fait

X\_train (675, 21325)

y train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.883

meilleur estimateur SVC(C=5, random\_state=42)

Accuracy: 0.911

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.92500 0.89888	0.89157 0.93023	0.90798 0.91429	83 86
accuracy macro avg weighted avg	0.91194 0.91171	0.91090 0.91124	0.91124 0.91113 0.91119	169 169 169

Ensemble des meilleurs paramètres :

C: 5

gamma: 'scale' Kernel: 'rbf' grid search fait

X\_train (675, 21197)

y\_train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits

```
meilleur score 0.884
```

meilleur estimateur SVC(C=5, random\_state=42)

Accuracy: 0.905

Classification Report

Ctassiiicatio	ni Neboi c			
	precision	recall	f1-score	support
FALSE	0.90361	0.90361	0.90361	83
TRUE	0.90698	0.90698	0.90698	86
accuracy			0.90533	169
macro avg	0.90530	0.90530	0.90530	169
weighted avg	0.90533	0.90533	0.90533	169

Ensemble des meilleurs paramètres :

C: 5

gamma: 'scale' kernel: 'rbf'

grid search fait X\_train (675, 14382)

y\_train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.883

meilleur estimateur SVC(C=2, random\_state=42)

Accuracy: 0.905

Classification Report

support	f1-score	recall	precision	
83 86	0.90123 0.90909	0.87952 0.93023	0.92405 0.88889	FALSE TRUE
169 169 169	0.90533 0.90516 0.90523	0.90488 0.90533	0.90647 0.90616	accuracy macro avg weighted avg

Ensemble des meilleurs paramètres :

C: 2

gamma: 'scale' kernel: 'rbf' grid search fait

X train (675, 25358)

y\_train (675,)

Fitting 5 folds for each of 18 candidates, totalling 90 fits meilleur score 0.877

meilleur estimateur SVC(C=2, random\_state=42)

Accuracy: 0.899

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.90244 0.89655	0.89157 0.90698	0.89697 0.90173	83 86
accuracy macro avg weighted avg	0.89950 0.89944	0.89927 0.89941	0.89941 0.89935 0.89939	169 169 169

Ensemble des meilleurs paramètres :

gamma: 'scale'
kernel: 'rbf'

grid search fait

X train (675, 25358)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.856

meilleur estimateur RandomForestClassifier(n estimators=50, random\_state=42)

Accuracy: 0.882

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.89873 0.86667	0.85542 0.90698	0.87654 0.88636	83 86
accuracy macro avg weighted avg	0.88270 0.88242	0.88120 0.88166	0.88166 0.88145 0.88154	169 169 169

Ensemble des meilleurs paramètres :

n estimators: 50

max features: 'sqrt'

grid search fait

X\_train (675, 21325)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.852

meilleur estimateur RandomForestClassifier(random\_state=42)

Accuracy: 0.882

Classification Report

	precision	recall	f1-score	support	
FALSE TRUE	0.89873 0.86667			83 86	
accuracy macro avg weighted avg		0.88120 0.88166		169 169 169	
Ensemble des meilleurs paramètres :     n_estimators: 100     max_features: 'sqrt' grid search fait X_train (675, 21197) y_train (675,) Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.855					
meilleur esti random_state=		omForestCl	assifier(n_	estimators=50	Э,
Accuracy : 0. Classificatio		recall	f1-score	support	
FALSE TRUE		0.86747 0.90698		83 86	
accuracy macro avg weighted avg				169 169 169	
<pre>Ensemble des meilleurs paramètres :</pre>					
<pre>meilleur estimateur RandomForestClassifier(max_features='log2', random_state=42)</pre>					
Accuracy : 0. Classificatio		recall	f1-score	support	
FALSE	0.84270	0.90361	0.87209	83	

TRU	JE	0.90000	0.83721	0.86747	86		
accurac macro av weighted av	/g		0.87041 0.86982		169 169 169		
Ensemble des meilleurs paramètres :     n_estimators: 100     max_features: 'log2' grid search fait X_train (675, 21325) y_train (675,) Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.855							
meilleur es random_stat			mForestCla	assifier(ma	x_features='log2'	,	
Accuracy : Classificat	ion		recall	fl-score	support		
	SE JE		0.93976 0.83721	0.89143 0.88344	83 86		
accurac macro av weighted av	/g	0.89145 0.89222	0.88848 0.88757		169 169 169		
<pre>Ensemble des meilleurs paramètres :</pre>							
<pre>meilleur estimateur RandomForestClassifier(random_state=42)</pre>							
Accuracy: 0.911 Classification Report     precision recall f1-score support							
FALS TRU		0.90476 0.91765	0.91566 0.90698	0.91018 0.91228	83 86		
accurac macro av	-	0.91120	0.91132	0.91124 0.91123	169 169		

Ensemble des meilleurs paramètres :

n\_estimators: 100
max features: 'sqrt'

grid search fait

X train (675, 14382)

y train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.864

meilleur estimateur RandomForestClassifier(n\_estimators=300, random state=42)

Accuracy: 0.882

Classification Report

	precision	recall	f1-score	support
FALSE TRUE	0.86207 0.90244	0.90361 0.86047	0.88235 0.88095	83 86
accuracy macro avg weighted avg	0.88225 0.88261	0.88204 0.88166	0.88166 0.88165 0.88164	169 169 169

Ensemble des meilleurs paramètres :

n\_estimators: 300

max features: 'sqrt'

grid search fait

X train (675, 25358)

y\_train (675,)

Fitting 5 folds for each of 8 candidates, totalling 40 fits meilleur score 0.870

meilleur estimateur RandomForestClassifier(n\_estimators=300, random\_state=42)

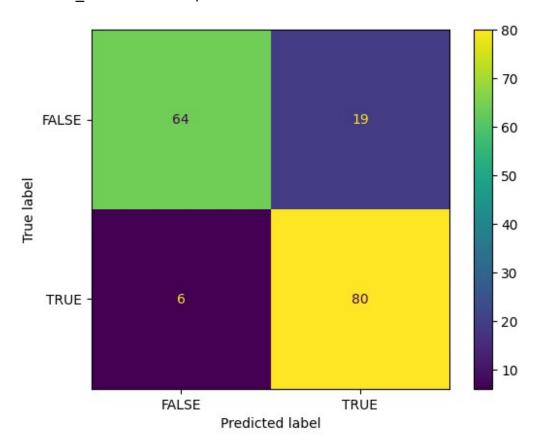
Accuracy: 0.899

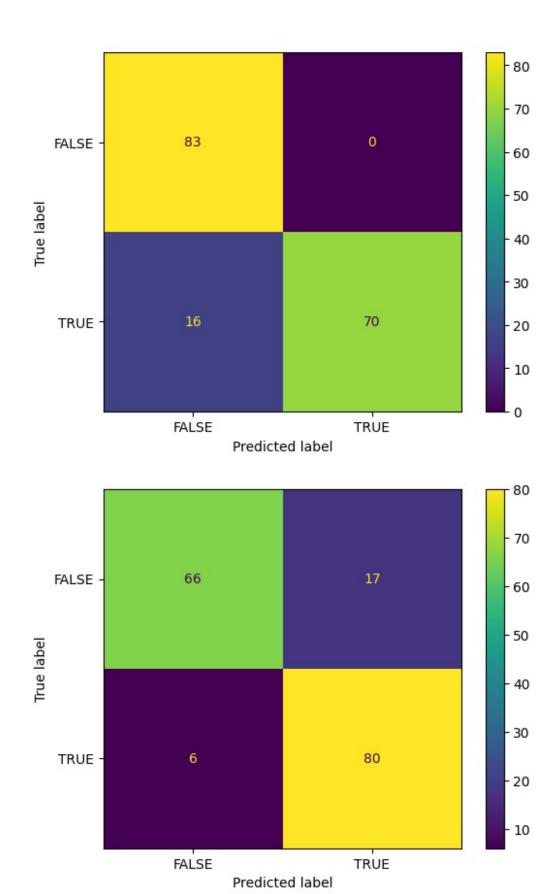
Classification Report

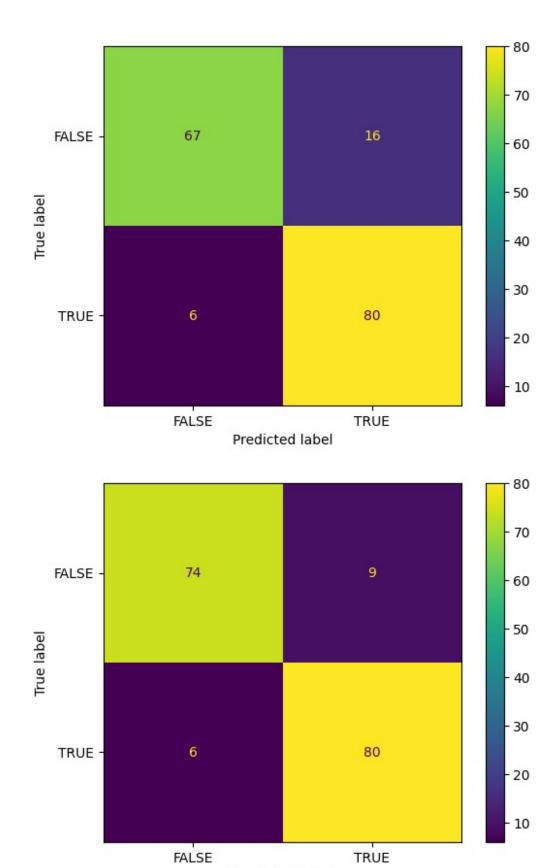
	precision	recall	f1-score	support
FALSE TRUE	0.88372 0.91566	0.91566 0.88372	0.89941 0.89941	83 86
accuracy macro avg weighted avg	0.89969 0.89998	0.89969 0.89941	0.89941 0.89941 0.89941	169 169 169

Ensemble des meilleurs paramètres :

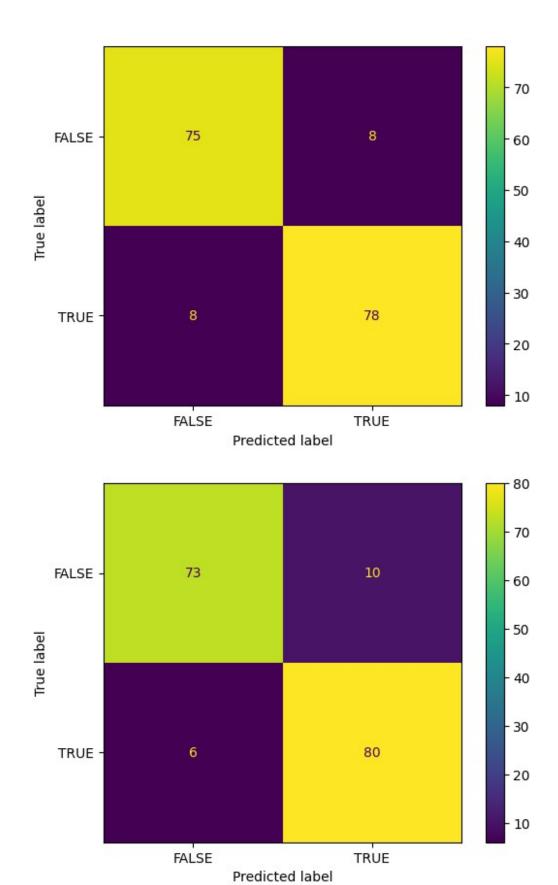
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max\_features: 'sqrt'

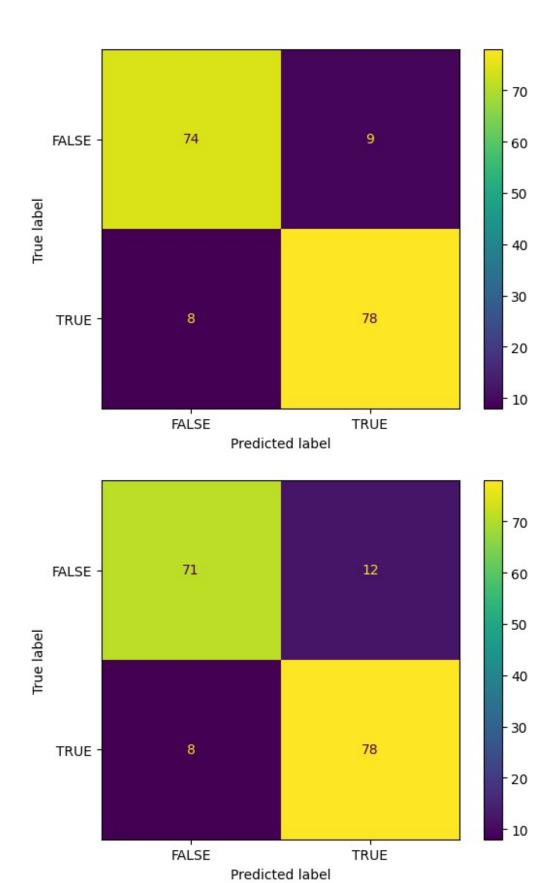


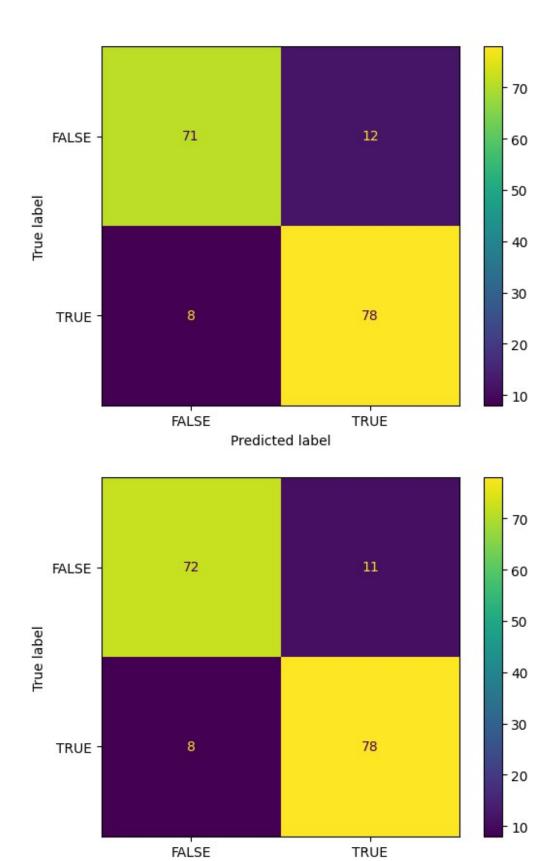




Predicted label







Predicted label

