Objective - The main objective of this project is to build a text classifier using CNN which takes text as the input and predicts the label.

Dataset details – The dataset contains 20,000 (Approx) newsgroup documents from 20 different newsgroups. The dataset is of size 46 MB approximately.

- Data source https://archive.ics.uci.edu/dataset/113/twenty+newsgroups)
- Data size (MB, GB, etc.) 46 MB approx.
- Data shape (# of rows and # columns) It does not contain data in the form of rows and columns rather data is present in the form text files (.txt format)

Research Question – Convolution Neural Networks are usually used for image processing. So, for this study I am going to analyze if it can also be used for processing text.

Approach

- The data here contains in the form of text files and does not contain any tabular format data.
- The text documents are first cleaned and processed using python regex which involves removing punctuations, converting non-English words to English etc.,
- The cleaned text documents are then vectorized.
- We use pre-trained glove vectors for vectorizing the text.
- CNN architectures are built and are trained using the above vectorized documents.
- · Results are evaluated for the train, test datasets.

```
In [1]:
```

- 1 # Spacy is a natural processing Language Library in Python which provides
- 2 # various languages.
- 3 # Installing Spacy
- 4 !pip install spacy

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Requirement already satisfied: spacy in /usr/local/lib/python3.10/dist-packag
es (3.6.1)
Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.11 in /usr/local/lib/
python3.10/dist-packages (from spacy) (3.0.12)
Requirement already satisfied: spacy-loggers<2.0.0,>=1.0.0 in /usr/local/lib/
python3.10/dist-packages (from spacy) (1.0.5)
Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /usr/local/lib/py
thon3.10/dist-packages (from spacy) (1.0.10)
Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.
10/dist-packages (from spacy) (2.0.8)
Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python
3.10/dist-packages (from spacy) (3.0.9)
Requirement already satisfied: thinc<8.2.0,>=8.1.8 in /usr/local/lib/python3.
10/dist-packages (from spacy) (8.1.12)
Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in /usr/local/lib/python
3.10/dist-packages (from spacy) (1.1.2)
Requirement already satisfied: srsly<3.0.0,>=2.4.3 in /usr/local/lib/python3.
10/dist-packages (from spacy) (2.4.8)
Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in /usr/local/lib/pyth
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Requirement already satisfied: typer<0.10.0,>=0.3.0 in /usr/local/lib/python
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Requirement already satisfied: pathy>=0.10.0 in /usr/local/lib/python3.10/dis
t-packages (from spacy) (0.10.3)
Requirement already satisfied: smart-open<7.0.0,>=5.2.1 in /usr/local/lib/pyt
hon3.10/dist-packages (from spacy) (6.4.0)
Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.
10/dist-packages (from spacy) (4.66.1)
Requirement already satisfied: numpy>=1.15.0 in /usr/local/lib/python3.10/dis
t-packages (from spacy) (1.23.5)
Requirement already satisfied: requests<3.0.0,>=2.13.0 in /usr/local/lib/pyth
on3.10/dist-packages (from spacy) (2.31.0)
Requirement already satisfied: pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4 in /usr/l
ocal/lib/python3.10/dist-packages (from spacy) (1.10.13)
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ges (from spacy) (3.1.2)
Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-p
ackages (from spacy) (67.7.2)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/d
ist-packages (from spacy) (23.2)
Requirement already satisfied: langcodes<4.0.0,>=3.2.0 in /usr/local/lib/pyth
on3.10/dist-packages (from spacy) (3.3.0)
Requirement already satisfied: typing-extensions>=4.2.0 in /usr/local/lib/pyt
hon3.10/dist-packages (from pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4->spacy) (4.
5.0)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/pyt
hon3.10/dist-packages (from requests<3.0.0,>=2.13.0->spacy) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist
-packages (from requests<3.0.0,>=2.13.0->spacy) (3.6)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.1
0/dist-packages (from requests<3.0.0,>=2.13.0->spacy) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.1
0/dist-packages (from requests<3.0.0,>=2.13.0->spacy) (2023.11.17)
Requirement already satisfied: blis<0.8.0,>=0.7.8 in /usr/local/lib/python3.1
0/dist-packages (from thinc<8.2.0,>=8.1.8->spacy) (0.7.11)
Requirement already satisfied: confection<1.0.0,>=0.0.1 in /usr/local/lib/pyt
hon3.10/dist-packages (from thinc<8.2.0,>=8.1.8->spacy) (0.1.4)
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Requirement already satisfied: click<9.0.0,>=7.1.1 in /usr/local/lib/python3. 10/dist-packages (from typer<0.10.0,>=0.3.0->spacy) (8.1.7)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/d ist-packages (from jinja2->spacy) (2.1.3)

```
2023-12-15 04:16:02.477913: E external/local xla/xla/stream executor/cuda/cud
a dnn.cc:9261] Unable to register cuDNN factory: Attempting to register facto
ry for plugin cuDNN when one has already been registered
2023-12-15 04:16:02.477970: E external/local_xla/xla/stream_executor/cuda/cud
a_fft.cc:607] Unable to register cuFFT factory: Attempting to register factor
y for plugin cuFFT when one has already been registered
2023-12-15 04:16:02.479883: E external/local_xla/xla/stream_executor/cuda/cud
a blas.cc:1515] Unable to register cuBLAS factory: Attempting to register fac
tory for plugin cuBLAS when one has already been registered
2023-12-15 04:16:02.491176: I tensorflow/core/platform/cpu_feature_guard.cc:1
82] This TensorFlow binary is optimized to use available CPU instructions in
performance-critical operations.
To enable the following instructions: AVX2 FMA, in other operations, rebuild
TensorFlow with the appropriate compiler flags.
2023-12-15 04:16:04.188699: W tensorflow/compiler/tf2tensorrt/utils/py_utils.
cc:38] TF-TRT Warning: Could not find TensorRT
2023-12-15 04:16:07.067587: I external/local_xla/xla/stream_executor/cuda/cud
a_executor.cc:901] successful NUMA node read from SysFS had negative value (-
1), but there must be at least one NUMA node, so returning NUMA node zero. Se
e more at https://github.com/torvalds/linux/blob/v6.0/Documentation/ABI/testi
ng/sysfs-bus-pci#L344-L355 (https://github.com/torvalds/linux/blob/v6.0/Docum
entation/ABI/testing/sysfs-bus-pci#L344-L355)
2023-12-15 04:16:07.068201: I external/local_xla/xla/stream_executor/cuda/cud
a executor.cc:901] successful NUMA node read from SysFS had negative value (-
1), but there must be at least one NUMA node, so returning NUMA node zero. Se
e more at https://github.com/torvalds/linux/blob/v6.0/Documentation/ABI/testi
ng/sysfs-bus-pci#L344-L355 (https://github.com/torvalds/linux/blob/v6.0/Docum
entation/ABI/testing/sysfs-bus-pci#L344-L355)
2023-12-15 04:16:07.068482: I external/local_xla/xla/stream_executor/cuda/cud
a_executor.cc:901] successful NUMA node read from SysFS had negative value (-
1), but there must be at least one NUMA node, so returning NUMA node zero. Se
e more at https://github.com/torvalds/linux/blob/v6.0/Documentation/ABI/testi
ng/sysfs-bus-pci#L344-L355 (https://github.com/torvalds/linux/blob/v6.0/Docum
entation/ABI/testing/sysfs-bus-pci#L344-L355)
⚠ As of spaCy v3.0, shortcuts like 'en' are deprecated. Please use the
full pipeline package name 'en_core_web_sm' instead.
Collecting en-core-web-sm==3.6.0
 Downloading https://github.com/explosion/spacy-models/releases/download/en
core web sm-3.6.0/en core web sm-3.6.0-py3-none-any.whl (https://github.com/e
xplosion/spacy-models/releases/download/en_core_web_sm-3.6.0/en_core_web_sm-
3.6.0-py3-none-any.whl) (12.8 MB)
                                            - 12.8/12.8 MB 21.4 MB/s eta 0:0
0:00
Requirement already satisfied: spacy<3.7.0,>=3.6.0 in /usr/local/lib/python3.
10/dist-packages (from en-core-web-sm==3.6.0) (3.6.1)
Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.11 in /usr/local/lib/
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thon3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (1.
0.10)
Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.
10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.0.8)
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Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python

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3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (3.0.9)
Requirement already satisfied: thinc<8.2.0,>=8.1.8 in /usr/local/lib/python3.
10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (8.1.12)
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Requirement already satisfied: srsly<3.0.0,>=2.4.3 in /usr/local/lib/python3.
10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.4.8)
Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in /usr/local/lib/pyth
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Requirement already satisfied: typer<0.10.0,>=0.3.0 in /usr/local/lib/python
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Requirement already satisfied: pathy>=0.10.0 in /usr/local/lib/python3.10/dis
t-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (0.10.3)
Requirement already satisfied: smart-open<7.0.0,>=5.2.1 in /usr/local/lib/pyt
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Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.
10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (4.66.1)
Requirement already satisfied: numpy>=1.15.0 in /usr/local/lib/python3.10/dis
t-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (1.23.5)
Requirement already satisfied: requests<3.0.0,>=2.13.0 in /usr/local/lib/pyth
on3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.31.
Requirement already satisfied: pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4 in /usr/l
ocal/lib/python3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==
3.6.0) (1.10.13)
Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packa
ges (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (3.1.2)
Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-p
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Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/d
ist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (23.2)
Requirement already satisfied: langcodes<4.0.0,>=3.2.0 in /usr/local/lib/pyth
on3.10/dist-packages (from spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (3.3.
0)
Requirement already satisfied: typing-extensions>=4.2.0 in /usr/local/lib/pyt
hon3.10/dist-packages (from pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4->spacy<3.7.
0, >= 3.6.0 - \text{en-core-web-sm} == 3.6.0) (4.5.0)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/pyt
hon3.10/dist-packages (from requests<3.0.0,>=2.13.0->spacy<3.7.0,>=3.6.0->en-
core-web-sm==3.6.0) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist
-packages (from requests<3.0.0,>=2.13.0->spacy<3.7.0,>=3.6.0->en-core-web-sm=
=3.6.0)(3.6)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.1
0/dist-packages (from requests<3.0.0,>=2.13.0->spacy<3.7.0,>=3.6.0->en-core-w
eb-sm==3.6.0) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.1
0/dist-packages (from requests<3.0.0,>=2.13.0->spacy<3.7.0,>=3.6.0->en-core-w
eb-sm==3.6.0) (2023.11.17)
Requirement already satisfied: blis<0.8.0,>=0.7.8 in /usr/local/lib/python3.1
0/dist-packages (from thinc<8.2.0,>=8.1.8->spacy<3.7.0,>=3.6.0->en-core-web-s
m==3.6.0) (0.7.11)
Requirement already satisfied: confection<1.0.0,>=0.0.1 in /usr/local/lib/pyt
hon3.10/dist-packages (from thinc<8.2.0,>=8.1.8->spacy<3.7.0,>=3.6.0->en-core
-web-sm==3.6.0) (0.1.4)
```

```
Requirement already satisfied: click<9.0.0,>=7.1.1 in /usr/local/lib/python3.
        10/dist-packages (from typer<0.10.0,>=0.3.0->spacy<3.7.0,>=3.6.0->en-core-web
        -sm = 3.6.0) (8.1.7)
        Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/d
        ist-packages (from jinja2->spacy<3.7.0,>=3.6.0->en-core-web-sm==3.6.0) (2.1.
        3)
        ✓ Download and installation successful
        You can now load the package via spacy.load('en_core_web_sm')
In [3]:
            import zipfile
          2 import re
          3 import spacy
          4 from tadm import tadm
In [4]:
          1 from google.colab import drive
          2 drive.mount('/content/drive')
        Mounted at /content/drive
In [5]:
            #We have dataset in the form of zip file
            #Unzipping the zip file.
          4 z = zipfile.ZipFile('/content/drive/MyDrive/Srisailam/20_newsgroups.zip')
          5 z.extractall()
In [6]:
            #extracting the class labels from the file paths to check its value counts
          2
            import os
          3
          4 | dir labels = []
          5 dir paths = []
            for root, dirs, files in os.walk("20_newsgroups", topdown=False):
          7
                for name in files:
                    x = os.path.join(root, name)
          8
          9
                    dir_labels.append(x.split('/')[-2])
         10
                    dir paths.append(x)
```

```
In [7]: 1 len(dir_labels)
```

Out[7]: 19997

There are 19997 documents in the dataset.

```
In [8]:
             import numpy as np
          2 list(np.unique(np.array(dir_labels)))
Out[8]: ['alt.atheism',
          'comp.graphics',
          'comp.os.ms-windows.misc',
          'comp.sys.ibm.pc.hardware',
          'comp.sys.mac.hardware',
          'comp.windows.x',
          'misc.forsale',
          'rec.autos',
          'rec.motorcycles',
          'rec.sport.baseball',
          'rec.sport.hockey',
          'sci.crypt',
          'sci.electronics',
          'sci.med',
          'sci.space',
          'soc.religion.christian',
          'talk.politics.guns',
          'talk.politics.mideast',
          'talk.politics.misc',
          'talk.religion.misc']
```

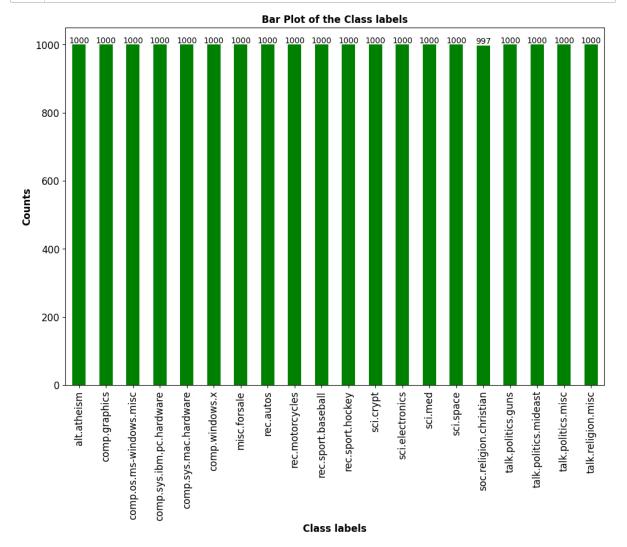
There are a total of 20 unique class labels as shown above.

Exploratory Data Analysis:

Figure 1:

```
In [9]: 1 import matplotlib.pyplot as plt
2 import pandas as pd
```

```
In [10]:
              #Bar chart showing the counts of each class label
           1
           2
              plt.figure(figsize=(12, 8))
           3
           4
             plt.xticks(fontsize=12)
              plt.yticks(fontsize=12)
           5
              ax = pd.Series(dir_labels).value_counts().sort_index().plot(kind='bar',col
           7
           8
              for c in ax.containers:
           9
                  # Optional: if the segment is small or 0, customize the labels
          10
                  labels = [v.get_height() if v.get_height() > 0 else '' for v in c]
                  # remove the labels parameter if it's not needed for customized labels
          11
                  ax.bar_label(c, labels=labels, label_type='edge')
          12
          13
          14 plt.xticks(rotation = 90)
             plt.title("Bar Plot of the Class labels",fontweight='bold')
          15
          16 | plt.ylabel("Counts", fontsize=12, fontweight='bold')
              plt.xlabel("Class labels",fontsize=12,fontweight='bold')
          17
              plt.show()
          18
```



Observations:

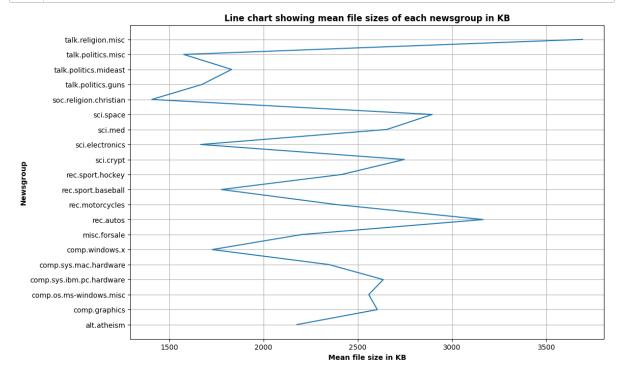
From the above bar chart it is clear that

- All the class labels has equal number of documents except for the soc.religion.christian class label.
- soc.religion.christian class label has 997 documents while all the other class labels have 1000 documents.

Figure 2:

```
In [11]:
              # Line chart showing the mean file size of each newsgroup.
           2
           3
             file_size = {}
              for root, dirs, files in os.walk("20_newsgroups", topdown=False):
           4
           5
                  for name in files:
                      x = os.path.join(root, name)
           6
           7
                      group_label = x.split('/')[-2]
           8
                      if group_label in file_size:
           9
                          file_size[group_label].append(os.path.getsize(x))
          10
                      else:
                          file_size[group_label] = [os.path.getsize(x)]
          11
```

```
In [12]: 1 file_sizes_mean = [np.array(file_size[i]).mean() for i in file_size.keys()
```



Observations:

From the above line chart it can be observed that the

- talk.religion.misc has got the highest mean file size.
- soc.religion.christian has got the lowest mean file size.

Progress for Presentation - 2

Data Cleaning and Preparation

Name of the article - 20_newsgroups/comp.graphics/38937

```
Path: cantaloupe.srv.cs.cmu.edu!magnesium.club.cc.cmu.edu!pitt.edu!uunet!mdis
ea!mothost!lmpsbbs!davidla@ecs.comm.mot.com
From: davidla@ecs.comm.mot.com (David Lau)
Newsgroups: comp.graphics
Subject: GIF viewer source code for PC?
Message-ID: <1993Apr29.152235.22224@lmpsbbs.comm.mot.com>
Date: 29 Apr 93 15:22:35 GMT
Sender: news@lmpsbbs.comm.mot.com (Net News)
Organization: Motorola
Lines: 10
Nntp-Posting-Host: 145.1.161.169
```

Does anyone know where I could get some source code for a GIF viewer on a PC. It doesn't have to be fancy, it would be best if it were written in C to display gif pictures on a VGA screen. I was hoping to get it for free from an FTP site, but I haven't been able to find it yet.

```
David Lau davidla@ecs.comm.mot.com
```

```
In [15]:
              def emailprocessing(input_text):
           1
           2
           3
                  This function takes the text as the input in
           4
                  string format and processes the words with
           5
                  email extensions and returns a string which
                  is concatenated with list of such words.
           6
           7
           8
                  raw_text = input_text.read()
           9
                  emails = re.findall(r"[\w\-\.]+\@[\w\.-]+\b", raw_text)# extracts email
          10
                  processed email
                  for i in emails:
          11
                      i = i.split('@')[1] # split the word based on '@'
          12
          13
                      temp = i.split('.')
          14
                      for j in temp:
          15
                          if len(j)<=2:# check for the Length of the word if it is Less</pre>
          16
          17
                          elif j=='com':# check if the word is 'com' and ignore if its t
                               continue
          18
          19
                          elif "@" in j:
          20
                               processed_email.append(j[1:])
          21
                          else:
          22
                               processed_email.append(j)
                  return ' '.join(processed_email).lower()
          23
```

ecs comm mot ecs comm mot lmpsbbs comm mot lmpsbbs comm mot ecs comm mot

```
In [17]:
           1
              def subjectprocessing(input_text):
           2
           3
                  This function takes the text as the input and
           4
                  processes the text line after the word "Subject:"
           5
                  and returns the string.
           6
           7
                  raw_text = input_text.read()
                  subject = re.findall('^Subject.*$', raw_text,re.MULTILINE) # extracts
           8
           9
                                                                              # that star
          10
                                                                              # as list d
          11
                  processed_subject = []
          12
                  subject = re.sub('[^A-Za-z0-9]+', ' ', subject[0]) # Replace non alpha
          13
                                                                      # single space.
                  subject = subject.rstrip().split(' ')# removes the leading and trailir
          14
          15
                                                        # and splits the text into words
                  for i in subject:
          16
          17
                      if i=='Subject' or i=='Re':# checks if the word is "Subject" or "F
                          continue
          18
          19
                      else:
                          processed_subject.append(i)
          20
          21
                  return ' '.join(processed_subject).lower()# joins the list of words in
              # Applying subjectprocessing function for the above text in the article
In [18]:
             # and printing its output
           3
           4 | file = open(dir_paths[1],encoding="utf8", errors='ignore')
           5 print(subjectprocessing(file))
           6 file.close()
```

gif viewer source code for pc

```
In [19]:
              def textprocessing(input text):
           2
           3
                  raw_text = input_text.read()
           4
                  text = re.sub(r"[\w\-\.]+@[\w\.-]+\b",' ',raw_text) # finds all occu
           5
                                                                         # and replaces t
                 text = re.sub(r'^Subject.*$',' ',text,flags=re.MULTILINE) # finds al
           6
           7
                                                                               # that sta
           8
                                                                               # a single
                          re.sub(r'Write to:.*$','', text,flags=re.MULTILINE) # replace
           9
                 text
          10
                                                                                # an empt
                          re.sub(r'From:.*$','', text,flags=re.MULTILINE)
          11
                                                                               # replace
                 text
          12
                                                                                # an empt
          13
                  #Decontractions
                  text = re.sub("[a-zA-Z]+:",'',text) # finds all occurrences of seque
          14
          15
                                                        # followed by a colon, and repla
          16
                       = re.sub(r"won\'t", "will not", text) # replaces the words wit
                  text
                 text = re.sub(r"can\'t", "can not", text)
          17
                                                                # replaces the words wit
                  text = re.sub(r"n\'t", " not", text)
          18
                                                                # replaces the words wit
                       = re.sub(r"\'re",
                                          , " are", text)
          19
                  text
                                                                # replaces the words wit
                          re.sub(r"\'s", " is", text)
          20
                                                                # replaces the words wit
                  text =
                       = re.sub(r"\'d", " would", text)
          21
                                                                # replaces the words wit
                  text
                      = re.sub(r"\'ll", " will", text)
          22
                  text
                                                                # replaces the words wit
                 text = re.sub(r"\'t", " not", text)
          23
                                                                # replaces the words wit
                 text = re.sub(r"\'ve", " have", text)
          24
                                                                # replaces the words wit
                       = re.sub(r"\'m", " am", text)
          25
                 text
                                                                # replaces the words wit
          26
                          text.strip()
                                                                # remove the leading and
                  text
          27
                       = re.sub(r'\<[^<>]*\>','', text,flags=re.MULTILINE) # removes t
                  text
          28
                                                                              # and does
                 text = re.sub(r"\([^()]*\)",'', text,flags=re.MULTILINE) # removes t
          29
          30
                                                                              # and does
                  text = re.sub(r'\[[^[]*\]','', text,flags=re.MULTILINE)
          31
                                                                              # removes 1
          32
                                                                              # and does
                  text = text.replace('>','') # replaces the parenthesis > with an en
          33
                          text.replace('\\','
                                             '') # replaces the backslash with an empty
          34
                  text
                          text.replace('?',' ') # replaces the '?' with an empty string
          35
                  text
          36
                       = re.sub('\n',' ', text,flags=re.MULTILINE) # replaces all the
                  text
          37
                                                                      # and does this for
                  text = re.sub(' +',' ',text,flags=re.MULTILINE) # replaces all the
          38
          39
                                                                      # and does this for
                  text = re.sub('-',' ', text,flags=re.MULTILINE) # replaces multiple
          40
          41
                                                                      # and does this for
          42
                  # Spacy is a natural processing language library in Python which provi
          43
                  # various languages including english.
          44
                  NER = spacy.load("en_core_web_sm") # loads the english language model.
          45
                  text1= NER(text) # text is tokenized and further linguistic analyses of
          46
                  a = []
          47
                  for i in text1.ents: # Loops through the named entities (e.g., names d
          48
                                       # from the tokens.
          49
                      a.append((i.text,i.label )) # appends the text of recognized named
          50
                  for i in a:
          51
                      if i[1]=='PERSON': # checks if the label of the named entity is a
          52
                          text = text.replace(i[0],'') # replace person entities in the
          53
                      elif i[1]=='GPE': # checks if the label of the named entity is ged
                          if " " in i[0]:
          54
                              temp = i[0].replace(' ','_')
          55
          56
                              text = text.replace(i[0],temp) # if the word is a geopolit
          57
                                                             # with an underscore to cor
```

```
text = re.sub('\d',' ',text) #replaces digits(0-9) with a single space
58
                                        text = re.sub(r'\b_+|_+\b', '', text) # finds the underscores that app
 59
60
                                                                                                                                                                                                                                       # or at the end or have one or n
                                                                                                                                                                                                                                       # and replaces with an empty str
61
                                       text = re.sub(r"b[a-zA-Z]{1}_{([a-zA-Z]+)",r"\1",text)} # It searches
62
63
                                                                                                                                                                                                                                                                                                                             # a single let
                                                                                                                                                                                                                                                                                                                             # followed by
64
65
                                                                                                                                                                                                                                                                                                                             # it replaces i
66
                                                                                                                                                                                                                                                                                                                             # the one or mc
                                       text = re.sub(r"\b[a-zA-Z]{2}_([a-zA-Z]+)",r"\1",text) # It searches f
67
68
                                                                                                                                                                                                                                                                                                                             # two alphabeti
69
                                                                                                                                                                                                                                                                                                                             # underscore, d
70
                                                                                                                                                                                                                                                                                                                             # it replaces i
71
                                                                                                                                                                                                                                                                                                                             # the one or mo
72
                                       text = text.lower()
73
                                       text = re.sub(r'\b\w{1,2}\b','',text) # It searches for words (sequential searches) | text 
74
                                                                                                                                                                                                                                             # that consist of one or two ch
75
                                      text = re.sub(r'\b\w{15,}\b','',text) # It searches for words (sequential sequential searches) | text = re.sub(r'\b\w{15,}\b','',text) # It searches for words (sequential sequential seq
                                                                                                                                                                                                                                             # that consist of 15 or more ch
76
                                      text = re.sub(r'[^a-zA-Z_]',' ',text) # It searches for characters th
77
                                                                                                                                                                                                                                             # (both uppercase and Lowercase
78
                                                                                                                                                                                                                                            # them with space characters.
79
80
                                       text = re.sub(r' {2,}',' ',text)
                                                                                                                                                                                                                                           # It searches for two or more of
                                                                                                                                                                                                                                            # and replaces them with a sing
81
82
                                        return text
```

```
In [20]: 1 # Applying textprocessing function for the above text in the article
2 # and printing its output
3
4 file = open(dir_paths[1],encoding="utf8", errors='ignore')
5 print(textprocessing(file))
6 file.close()
```

cantaloupe srv cmu edu magnesium club cmu edu pitt edu uunet mdisea mothost l mpsbbs comp graphics message apr gmt motorola nntp posting does anyone know w here could get some source code for gif viewer does not have fancy would best were written display pictures vga screen was hoping get for free from ftp sit e but have not been able find yet

```
In [21]:
              email = []
              for i in tqdm(dir_paths): # Looping through all the files and
           2
           3
                                         # applying email processing function and
           4
                                         # appending the output of the funtion to the
           5
                                         # email list.
           6
                  sample = open(i,encoding="utf8", errors='ignore')
           7
                  email.append(emailprocessing(sample))
           8
                  sample.close()
```

100%| 100%| 1000 | 19997/19997 [00:08<00:00, 2333.04it/s]

```
In [22]:
                subject = []
            1
            2
                for i in tqdm(dir_paths): # looping through all the files and
                                              # applying subject processing function and
            3
            4
                                              # appending the output of the funtion to the
            5
                                              # subject list.
                    sample = open(i,encoding="utf8", errors='ignore')
            6
            7
                    subject.append(subjectprocessing(sample))
            8
                    sample.close()
           100%
                            | 19997/19997 [00:02<00:00, 8235.66it/s]
 In [ ]:
                text = []
                for i in tqdm(dir_paths): # looping through all the files and
             2
                                              # applying text processing function and
             3
                                              # appending the output of the funtion to the
            4
            5
                                              # text list.
                    sample = open(i,encoding="utf8", errors='ignore')
            6
            7
                    text.append(textprocessing(sample))
                    sample.close()
            8
 In [ ]:
                #creating a dataframe out of email, subject, text and dir_labels lists
            1
            2
               data = pd.DataFrame(list(zip(email, subject, text, dir_labels)), columns = ['e
               data.to pickle("processed data.pkl") # storing the dataframe in a pickle |
                unpickled_df = pd.read_pickle('/content/drive/MyDrive/Srisailam/processed_
In [23]:
                unpickled df.head()
Out[23]:
                                 email
                                                        subject
                                                                                    text target label
                    mantis mantis mit edu
                                             alt atheism faq atheist
                                                                  cantaloupe srv cmu edu alt
           0
                                                                                           alt.atheism
                    mantis netcom mantis
                                                      resources
                                                                       alt atheism news al...
                    mantis mantis mit edu
                                         alt atheism fag introduction
                                                                  cantaloupe srv cmu edu alt
                                                                                           alt.atheism
            1
                    mantis mantis mantis
                                                      to atheism
                                                                       alt atheism news al...
                 dbstu1 tu-bs dbstu1 tu-bs
                                                                   alt atheism cantaloupe srv
           2
                                                    gospel dating
                                                                                           alt.atheism
                     ibr tu-bs dbstu1 tu-...
                                                                     cmu edu crabapple s...
                 mantis mantis kepler unh
                                                university violating
                                                                  cantaloupe srv cmu edu alt
           3
                                                                                           alt.atheism
                      edu kepler unh edu
                                         separation of church state
                                                                        alt politics usa ca...
               watson ibm com watson ibm
                                          soc motss et al princeton
                                                                  cantaloupe srv cmu edu alt
                                                                                           alt.atheism
                   austin ibm cbnewsl a...
                                            axes matching funds ...
                                                                       soc rec alt atheism...
In [24]:
                # extracting the length of the words for each row in the text
                # and assigning to new column by name text word count.
            2
             3
               unpickled_df['text_word_count'] = unpickled_df['text'].apply(lambda x:len(
```

In [25]:

1 # printing first 5 rows of the data.
2 unpickled_df.head()

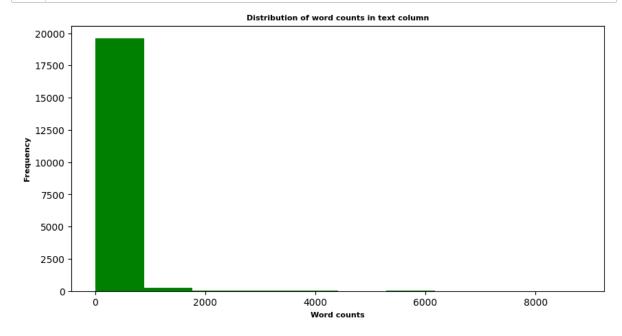
Out[25]:

	email	subject	text	target_label	text_word_count
0	mantis mantis mit edu mantis netcom mantis	alt atheism faq atheist resources	cantaloupe srv cmu edu alt alt atheism news al	alt.atheism	1224
1	mantis mantis mit edu mantis mantis mantis	alt atheism faq introduction to atheism	cantaloupe srv cmu edu alt alt atheism news al	alt.atheism	4023
2	dbstu1 tu-bs dbstu1 tu-bs ibr tu-bs dbstu1 tu	gospel dating	alt atheism cantaloupe srv cmu edu crabapple s	alt.atheism	518
3	mantis mantis kepler unh edu kepler unh edu	university violating separation of church state	cantaloupe srv cmu edu alt alt politics usa ca	alt.atheism	189
4	watson ibm com watson ibm austin ibm cbnewsl a	soc motss et al princeton axes matching funds	cantaloupe srv cmu edu alt soc rec alt atheism	alt.atheism	107

```
In [26]:
```

```
plt.figure(figsize=(10,5))
unpickled_df['text_word_count'].plot.hist(color='green')

plt.title("Distribution of word counts in text column",fontsize=8,fontweig
plt.xlabel("Word counts",fontsize=8,fontweight='bold')
plt.ylabel("Frequency",fontsize=8,fontweight='bold')
plt.show()
```

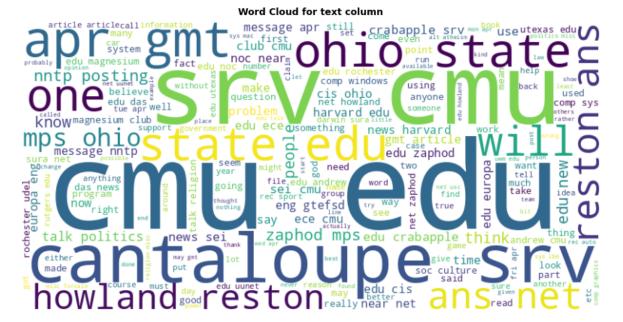


Observation:

From the above chart, it can be observed that the most of the word counts in the text column are less than 1000.

Displaying Wordclouds for subject and text columns

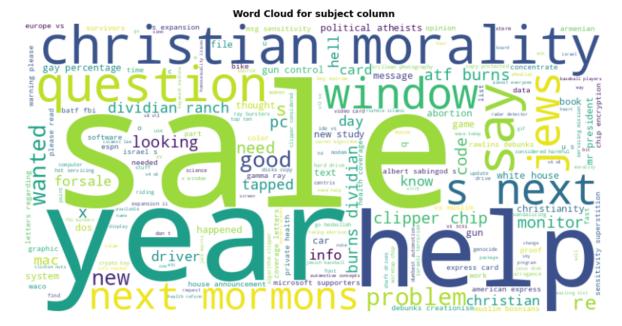
```
In [27]:
              from wordcloud import WordCloud
In [28]:
              #Word Cloud for text column
           2
             wordcloud = WordCloud(width=800, height=400, background_color="white").\
           3
           4
                          generate(" ".join(unpickled_df['text'].tolist()))
              # Display the word cloud using matplotlib
           5
              plt.figure(figsize=(10, 5))
              plt.imshow(wordcloud, interpolation="bilinear")
              plt.title("Word Cloud for text column",fontsize=9,fontweight='bold')
           8
           9
             plt.axis("off")
          10
             plt.show()
```



Observations:

It can be observed that the Ohio state is highly used word in the news article. News articles must have covered most of its news from Ohio state.

```
In [29]:
              # word cloud for subject column
           1
           2
             wordcloud = WordCloud(width=800, height=400, background_color="white").\
           3
                          generate(" ".join(unpickled_df['subject'].tolist()))
           4
              # Display the word cloud using matplotlib
           5
              plt.figure(figsize=(10, 5))
           7
              plt.imshow(wordcloud, interpolation="bilinear")
              plt.title("Word Cloud for subject column",fontsize=9,fontweight='bold')
              plt.axis("off")
          10
              plt.show()
```



Observations:

It can be observed that the sale and year are highly used words in the subject of news articles.

Train-Test Split

```
In [32]:
             # Splitting the data into train and test with test size of 25% of the total
          1
           2
           3 from sklearn.model_selection import train_test_split
           4 from sklearn import preprocessing
             import tensorflow as tf
           7 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25,
           8 le = preprocessing.LabelEncoder() # Encoding the target Labels using Label
          9 le.fit(y_train)
          10 y_train = le.transform(y_train)
          11 y test = le.transform(y test)
             y_train = tf.keras.utils.to_categorical(y_train) # converting labels in )
                      = tf.keras.utils.to_categorical(y_test) # converting labels in y
          13 y_test
In [33]:
             # printing the shapes of train and test datasets
          1
           2
           3 print(X_train.shape)
           4 print(X_test.shape)
           5 print(y_train.shape)
           6 print(y_test.shape)
         (14997,)
         (5000,)
         (14997, 20)
         (5000, 20)
In [34]:
          1 lengths = [len(s.split()) for s in X_train] # finding length of words in e
           2 max_length = int(np.percentile(lengths,99)) # finding the Length pertainir
           3 print(max length)
```

1361

The above max length gives information that 99% of the documents have less than or equal 1361 words in it.

Modelling

```
In [35]:
             import tensorflow
             from keras.preprocessing.text import Tokenizer
           3
             from keras.preprocessing.sequence import pad_sequences
           4
             tokenizer = Tokenizer(num words=10000, lower= 1,\
           5
                                    filters= '!"#$%&()*+,-./:;<=>?@[\\]^`{|}~\t\n',oov_t
           6
           7
                                                                    # the text data excl
           8
                                                                    # in the filters as
           9
             tokenizer.fit_on_texts(X_train)
             train_sequences = tokenizer.texts_to_sequences(X_train)
          10
             test sequences = tokenizer.texts to sequences(X test)
          11
          12
          13
             # The pad_sequences function pads or truncates the input sequences to have
          14
             # and it adds padding at the end of each sequence. The result is a set of
          15
          16 | train_padded = pad_sequences(train_sequences,padding="post",\
                                           truncating='post',maxlen=int(0.1*max_length),
          17
          18 test padded = pad sequences(test sequences,padding="post",\
          19
                                           truncating='post',maxlen=int(0.1*max_length),
          20
             # printing the shape of train and test datasets post padding the sequences
          21
          22 print(train padded.shape)
          23 print(test_padded.shape)
         (14997, 136)
         (5000, 136)
In [36]:
             size_of_vocabulary = len(tokenizer.word_index) + 1
             print(size of vocabulary)
         94450
In [38]:
             #Reference - https://www.analyticsvidhya.com/blog/2020/03/pretrained-word-
           1
           2
           3 # creating a dictionary(word:coeffs) using glove vectors text file.
           4 embeddings_index = dict()
           5 | f = open('/content/drive/MyDrive/Srisailam/glove.6B.100d.txt',encoding="ut
           6
             for line in f:
           7
                  values = line.split()
           8
                  word
                        = values[0]
                  coefs = np.asarray(values[1:],dtype='float32')
           9
          10
                  embeddings_index[word] = coefs
          11 f.close()
          12 print('Loaded {0} word vectors'.format(len(embeddings_index)))
```

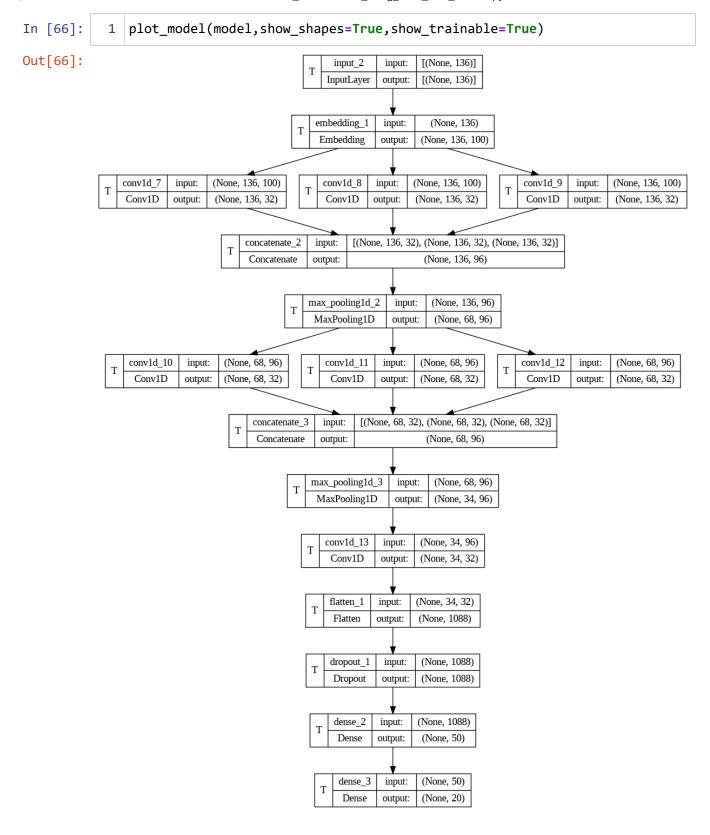
Loaded 400000 word vectors

```
In [39]:
              # finding the embedding matrix for the words in the tokenizer using
           1
              # embedding index created earlier.
           2
           3
           4
             embedding_matrix = np.zeros((size_of_vocabulary,100))
           5
             for word, i in tokenizer.word index.items():
           6
                  embedding vector = embeddings index.get(word)
           7
                  if embedding_vector is not None:
           8
                      embedding matrix[i] = embedding vector
```

CNN Architecture

```
In [64]:
             from keras.layers import Dense,Input,Activation,Embedding,Conv1D,concatend
                                     MaxPooling1D, Flatten, Dropout, BatchNormalization
           2
           3
             from keras.models import Model
             from keras.utils import plot model
           5
             initializer = tf.keras.initializers.GlorotNormal(seed=0)
             input_layer = Input(shape = (int(0.1*max_length),),dtype='int32')
           7
             word_embedding = Embedding(size_of_vocabulary,100,weights=[embedding_matri
           8
                                         input_length=1000,trainable=True)(input_layer)
           9
          10
             conv1
                           Conv1D(filters=32,kernel_size=4,activation='relu',padding='s@
                                  kernel_initializer=initializer)(word_embedding)
          11
                           Conv1D(filters=32,kernel size=6,activation='relu',padding='sa
          12 conv2
                                  kernel_initializer=initializer)(word_embedding)
          13
          14 conv3
                           Conv1D(filters=32,kernel_size=8,activation='relu',padding='sa
          15
                                  kernel initializer=initializer)(word embedding)
          16 merged1 =
                           concatenate([conv1,conv2,conv3])
                           MaxPooling1D(pool_size=2, padding='same')(merged1)
          17 pool1
                           Conv1D(filters=32, kernel_size=4, activation='relu', padding='sa
          18 conv4
                           Conv1D(filters=32, kernel_size=6, activation='relu', padding='sa
          19 conv5
          20 conv6
                           Conv1D(filters=32,kernel_size=8,activation='relu',padding='sa
                           concatenate([conv4,conv5,conv6])
          21 merged2 =
          22 pool2
                           MaxPooling1D(pool_size=2,padding='same')(merged2)
          23 conv7
                           Conv1D(filters=32,kernel_size=4,activation='relu',padding='se
          24 flat1
                           Flatten()(conv7)
                       =
          25 drop1
                           Dropout(0.5)(flat1)
                      =
                           Dense(50, activation='relu', kernel_initializer=initializer)(dr
          26 dense1
                           Dense(20,activation='softmax',kernel_initializer=initializer)
          27
             output
          28
          29 model
                           Model(inputs = input_layer,outputs= output)
In [65]:
             from keras.optimizers import Adam
```

```
2 adam = Adam(learning_rate=0.0001)
3 | model.compile(loss='categorical_crossentropy',optimizer = adam ,metrics=[
```



In [67]: 1 model.summary(show_trainable=True,positions=[0.35,0.6,0.7,1])

Model: "model_1"

Layer (type) to	Trainable	Output Shape	Param #	Connected
input_2 (InputLaye		[(None, 136)]	0	[]
embedding_1 (Embed	dding) Y	(None, 136, 100)	9445000	['input_2
conv1d_7 (Conv1D) ng_1[0][0]']	Υ	(None, 136, 32)	12832	['embeddi
conv1d_8 (Conv1D) ng_1[0][0]']	Υ	(None, 136, 32)	19232	['embeddi
conv1d_9 (Conv1D) ng_1[0][0]']	Υ	(None, 136, 32)	25632	['embeddi
<pre>concatenate_2 (Cor 7[0][0]',</pre>	ncatenate) Y	(None, 136, 96)	0	['conv1d_
8[0][0]',				conv1d_
9[0][0]']				'conv1d_
<pre>max_pooling1d_2 (N nate_2[0][0]']</pre>	MaxPooling1D) Y	(None, 68, 96)	0	['concate
conv1d_10 (Conv1D) ling1d_2[0][0]']) Y	(None, 68, 32)	12320	['max_poo
conv1d_11 (Conv1D) ling1d_2[0][0]']) Y	(None, 68, 32)	18464	['max_poo
conv1d_12 (Conv1D) ling1d_2[0][0]']) Y	(None, 68, 32)	24608	['max_poo
concatenate_3 (Cor 10[0][0]',	ncatenate) Y	(None, 68, 96)	0	['conv1d_ 'conv1d_
11[0][0]',				_
12[0][0]']				'conv1d_
<pre>max_pooling1d_3 (Mate_3[0][0]']</pre>	MaxPooling1D) Y	(None, 34, 96)	0	['concate
conv1d_13 (Conv1D) ling1d_3[0][0]']) Y	(None, 34, 32)	12320	['max_poo
flatten_1 (Flatter 13[0][0]']	1) Y	(None, 1088)	0	['conv1d_

```
(None, 1088)
                                                                        ['flatten
dropout 1 (Dropout)
                      Υ
_1[0][0]']
dense_2 (Dense)
                                    (None, 50)
                                                             54450
                                                                        ['dropout
                      Υ
_1[0][0]']
dense_3 (Dense)
                                    (None, 20)
                                                                        ['dense_2
                                                             1020
[0][0]']
```

Total params: 9625878 (36.72 MB) Trainable params: 9625878 (36.72 MB)

Non-trainable params: 0 (0.00 Byte)

```
In [68]:
           1 | from sklearn.metrics import f1_score
           2 from keras.callbacks import ModelCheckpoint
           3 from keras.callbacks import EarlyStopping
             from keras.callbacks import ReduceLROnPlateau
           5
             es = EarlyStopping(monitor='val_accuracy', mode='max', verbose=1,patience=
           7
             mc = ModelCheckpoint('/content/drive/MyDrive/Srisailam/best_model_1.h5', n
           8
                                   mode='max', save_best_only=True, verbose=1)
             reduce_lr = ReduceLROnPlateau(monitor='val_accuracy', factor=0.8, patience
           9
          10
             callback_list = [es,mc,reduce_lr]
          11
```

Epoch 1: val_accuracy improved from -inf to 0.08520, saving model to /content/drive/MyDrive/Srisailam/best_model_1.h5

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103: Us erWarning: You are saving your model as an HDF5 file via `model.save()`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')`.

saving_api.save_model(

```
235/235 [================ ] - 24s 93ms/step - loss: 3.0017 - acc
uracy: 0.0573 - val_loss: 2.9627 - val_accuracy: 0.0852 - lr: 1.0000e-04
Epoch 2/25
0.1283
Epoch 2: val_accuracy improved from 0.08520 to 0.22560, saving model to /cont
ent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [=============== ] - 9s 37ms/step - loss: 2.8062 - accu
racy: 0.1283 - val_loss: 2.4359 - val_accuracy: 0.2256 - lr: 1.0000e-04
Epoch 3/25
0.2492
Epoch 3: val_accuracy improved from 0.22560 to 0.34860, saving model to /cont
ent/drive/MyDrive/Srisailam/best model 1.h5
235/235 [=============== ] - 7s 28ms/step - loss: 2.2481 - accu
racy: 0.2492 - val_loss: 1.9057 - val_accuracy: 0.3486 - lr: 1.0000e-04
Epoch 4/25
0.3508
Epoch 4: val accuracy improved from 0.34860 to 0.43920, saving model to /cont
ent/drive/MyDrive/Srisailam/best model 1.h5
235/235 [=============== ] - 5s 22ms/step - loss: 1.8566 - accu
racy: 0.3508 - val loss: 1.6200 - val accuracy: 0.4392 - lr: 1.0000e-04
Epoch 5/25
0.4667
Epoch 5: val_accuracy improved from 0.43920 to 0.57720, saving model to /cont
ent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [============== ] - 4s 18ms/step - loss: 1.5009 - accu
racy: 0.4667 - val_loss: 1.2483 - val_accuracy: 0.5772 - lr: 1.0000e-04
Epoch 6/25
0.5701
Epoch 6: val_accuracy improved from 0.57720 to 0.66300, saving model to /cont
ent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [============== ] - 5s 23ms/step - loss: 1.2160 - accu
racy: 0.5701 - val loss: 1.0157 - val accuracy: 0.6630 - lr: 1.0000e-04
Epoch 7/25
0.6489
Epoch 7: val_accuracy improved from 0.66300 to 0.71900, saving model to /cont
ent/drive/MyDrive/Srisailam/best model 1.h5
racy: 0.6490 - val_loss: 0.8333 - val_accuracy: 0.7190 - lr: 1.0000e-04
Epoch 8/25
Epoch 8: val_accuracy improved from 0.71900 to 0.77560, saving model to /cont
ent/drive/MyDrive/Srisailam/best model 1.h5
racy: 0.7173 - val_loss: 0.6868 - val_accuracy: 0.7756 - lr: 1.0000e-04
Epoch 9/25
0.7658
Epoch 9: val accuracy improved from 0.77560 to 0.81140, saving model to /cont
ent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [================ ] - 4s 17ms/step - loss: 0.6794 - accu
```

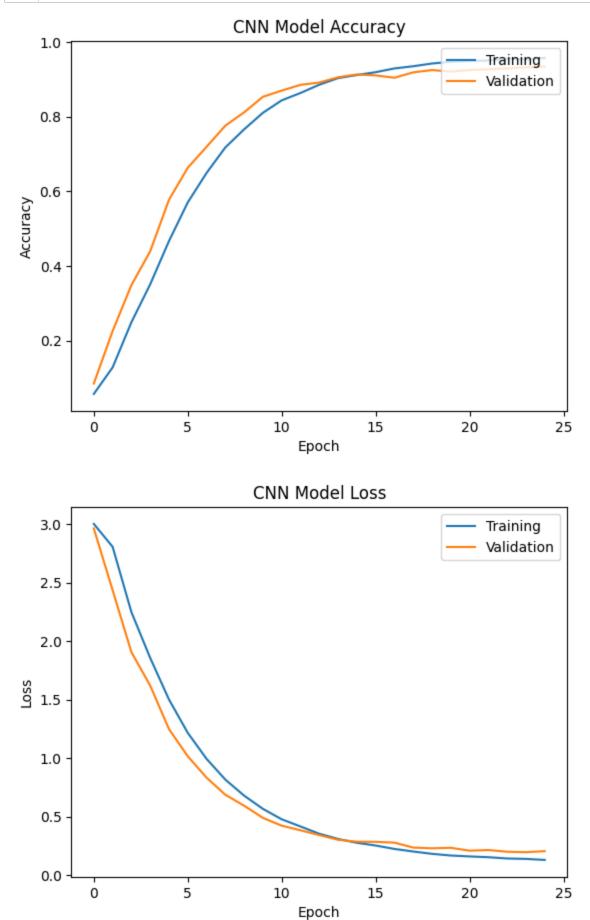
```
racy: 0.7658 - val loss: 0.5934 - val accuracy: 0.8114 - lr: 1.0000e-04
Epoch 10/25
0.8101
Epoch 10: val_accuracy improved from 0.81140 to 0.85300, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [=============== ] - 4s 17ms/step - loss: 0.5664 - accu
racy: 0.8101 - val loss: 0.4893 - val accuracy: 0.8530 - lr: 1.0000e-04
Epoch 11/25
235/235 [================== ] - ETA: 0s - loss: 0.4773 - accuracy:
0.8434
Epoch 11: val_accuracy improved from 0.85300 to 0.86960, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [============== ] - 3s 15ms/step - loss: 0.4773 - accu
racy: 0.8434 - val_loss: 0.4227 - val_accuracy: 0.8696 - lr: 1.0000e-04
Epoch 12/25
0.8632
Epoch 12: val_accuracy improved from 0.86960 to 0.88500, saving model to /con
tent/drive/MyDrive/Srisailam/best model 1.h5
235/235 [================ ] - 8s 36ms/step - loss: 0.4159 - accu
racy: 0.8632 - val_loss: 0.3833 - val_accuracy: 0.8850 - lr: 1.0000e-04
Epoch 13/25
0.8853
Epoch 13: val accuracy improved from 0.88500 to 0.89120, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [============== ] - 7s 30ms/step - loss: 0.3538 - accu
racy: 0.8853 - val_loss: 0.3425 - val_accuracy: 0.8912 - lr: 1.0000e-04
Epoch 14/25
235/235 [================== ] - ETA: 0s - loss: 0.3091 - accuracy:
0.9030
Epoch 14: val_accuracy improved from 0.89120 to 0.90520, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [=============== ] - 4s 16ms/step - loss: 0.3091 - accu
racy: 0.9030 - val_loss: 0.3022 - val_accuracy: 0.9052 - lr: 1.0000e-04
Epoch 15/25
235/235 [================= ] - ETA: 0s - loss: 0.2765 - accuracy:
0.9112
Epoch 15: val_accuracy improved from 0.90520 to 0.91280, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [================ ] - 8s 34ms/step - loss: 0.2765 - accu
racy: 0.9112 - val_loss: 0.2861 - val_accuracy: 0.9128 - lr: 1.0000e-04
Epoch 16/25
0.9189
Epoch 16: val_accuracy did not improve from 0.91280
racy: 0.9189 - val loss: 0.2845 - val accuracy: 0.9102 - lr: 1.0000e-04
Epoch 17/25
235/235 [============= ] - ETA: 0s - loss: 0.2239 - accuracy:
0.9289
Epoch 17: val_accuracy did not improve from 0.91280
racy: 0.9289 - val_loss: 0.2778 - val_accuracy: 0.9040 - lr: 1.0000e-04
Epoch 18/25
235/235 [================ ] - ETA: 0s - loss: 0.2019 - accuracy:
```

```
0.9348
Epoch 18: val accuracy improved from 0.91280 to 0.91840, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [================ ] - 5s 23ms/step - loss: 0.2019 - accu
racy: 0.9348 - val_loss: 0.2355 - val_accuracy: 0.9184 - lr: 8.0000e-05
Epoch 19/25
235/235 [================= ] - ETA: 0s - loss: 0.1820 - accuracy:
0.9421
Epoch 19: val_accuracy improved from 0.91840 to 0.92460, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [================ ] - 4s 16ms/step - loss: 0.1820 - accu
racy: 0.9421 - val_loss: 0.2294 - val_accuracy: 0.9246 - lr: 8.0000e-05
Epoch 20/25
0.9467
Epoch 20: val_accuracy did not improve from 0.92460
235/235 [=============] - 3s 12ms/step - loss: 0.1678 - accu
racy: 0.9467 - val_loss: 0.2337 - val_accuracy: 0.9204 - lr: 8.0000e-05
Epoch 21/25
0.9489
Epoch 21: val_accuracy improved from 0.92460 to 0.92480, saving model to /con
tent/drive/MyDrive/Srisailam/best model 1.h5
racy: 0.9489 - val_loss: 0.2090 - val_accuracy: 0.9248 - lr: 8.0000e-05
Epoch 22/25
0.9501
Epoch 22: val_accuracy improved from 0.92480 to 0.92660, saving model to /con
tent/drive/MyDrive/Srisailam/best model 1.h5
racy: 0.9501 - val_loss: 0.2148 - val_accuracy: 0.9266 - lr: 8.0000e-05
Epoch 23/25
0.9554
Epoch 23: val_accuracy improved from 0.92660 to 0.92900, saving model to /con
tent/drive/MyDrive/Srisailam/best model 1.h5
235/235 [=============== ] - 3s 13ms/step - loss: 0.1423 - accu
racy: 0.9555 - val_loss: 0.1998 - val_accuracy: 0.9290 - lr: 8.0000e-05
Epoch 24/25
Epoch 24: val_accuracy improved from 0.92900 to 0.93160, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
racy: 0.9530 - val_loss: 0.1962 - val_accuracy: 0.9316 - lr: 8.0000e-05
Epoch 25/25
Epoch 25: val_accuracy improved from 0.93160 to 0.93380, saving model to /con
tent/drive/MyDrive/Srisailam/best_model_1.h5
235/235 [============== ] - 3s 15ms/step - loss: 0.1300 - accu
racy: 0.9571 - val_loss: 0.2049 - val_accuracy: 0.9338 - lr: 8.0000e-05
```

Out[69]: <keras.src.callbacks.History at 0x78fd62b86320>

```
In [70]:
           1
              def plot_history(history,accuracy_title,loss_title):
           2
                  # summarize history for accuracy
           3
                  plt.plot(model.history.history['accuracy'])
           4
                  plt.plot(model.history.history['val_accuracy'])
           5
                  plt.title(accuracy title)
           6
                  plt.ylabel('Accuracy')
           7
                  plt.xlabel('Epoch')
                  plt.legend(['Training', 'Validation'], loc='upper right')
           8
           9
                  plt.show()
          10
                  # summarize history for loss
          11
                  plt.plot(model.history.history['loss'])
          12
          13
                  plt.plot(model.history.history['val_loss'])
          14
                  plt.title(loss_title)
                  plt.ylabel('Loss')
          15
                  plt.xlabel('Epoch')
          16
                  plt.legend(['Training', 'Validation'], loc='upper right')
          17
          18
                  plt.show()
```

In [71]: 1 plot_history(model,'CNN Model Accuracy','CNN Model Loss')



Machine Learning Models

```
In [72]:
          1 from sklearn.linear model import LogisticRegression
          2 from sklearn.ensemble import RandomForestClassifier
          3 from sklearn.metrics import accuracy_score
In [73]:
          1 LR = LogisticRegression(random_state=0, max_iter=10000).fit(train_padded, r
In [74]:
          1 train preds = LR.predict(train padded)
          2 train_accuracy = accuracy_score(np.argmax(y_train,axis=1), train_preds)
          3 print('Train Accuracy Score using Logistic Regression is ',train_accuracy)
          5 test_preds = LR.predict(test_padded)
          6 test_accuracy = accuracy_score(np.argmax(y_test,axis=1), test_preds)
             print('Test Accuracy Score using Logistic Regression is ',test_accuracy)
         Train Accuracy Score using Logistic Regression is 0.16309928652397146
         Test Accuracy Score using Logistic Regression is 0.1078
             RF = RandomForestClassifier(max_depth=4, random_state=0).fit(train_padded,
In [75]:
In [76]:
          1 train_preds = RF.predict(train_padded)
          2 train_accuracy = accuracy_score(np.argmax(y_train,axis=1), train_preds)
          3 print('Train Accuracy Score using Random Forest is ',train accuracy)
          5 test preds = RF.predict(test padded)
          6 test_accuracy = accuracy_score(np.argmax(y_test,axis=1), test_preds)
          7 print('Test Accuracy Score using Random Forest is ',test_accuracy)
         Train Accuracy Score using Random Forest is 0.15689804627592185
         Test Accuracy Score using Random Forest is 0.1286
In [80]:
          1 #Summarizing the results obtained from each trained model
          3 from prettytable import PrettyTable
          4 x = PrettyTable()
          5 | x.field_names = ["Model", "Train Accuracy", "Test Accuracy"]
          6 x.add_row(["CNN",0.957,0.933])
          7 x.add row(["Logistic Regression", 0.163, 0.10])
          8 | x.add_row(["Random Forest", 0.156, 0.128])
          9 print(x)
         +----+
                 Model | Train Accuracy | Test Accuracy |
                  CNN
                                   0.957
                                                    0.933
         | Logistic Regression |
                                   0.163
                                                     0.1
            Random Forest
                                   0.156
                                                    0.128
```

Observations:

We can observe that the CNN has outperformed both logistic regression and Random Forest in terms of accuracies.

```
In [78]: 1  from keras.models import load_model
2  best_model = load_model('/content/drive/MyDrive/Srisailam/best_model_1.h5'
```

```
In [79]:
              #Predictions on random 20 points in test data
           2
              import random
           3
             for _ in range(20):
                i = random.randint(0,X_test.shape[0])
           4
           5
                print(X_test[i])
                print('\n')
           6
           7
           8
                #Predicting the labels using the best model obtained above.
           9
                pred_label = np.argmax(best_model.predict(test_padded[i].reshape(1,-1)))
          10
                actual_label = np.argmax(y_test[i])
                print('Actual Label - ',le.classes_[actual_label])
          11
                print('Predicted Label - ',le.classes_[pred_label])
          12
                print(100*'*')
          13
```

descartes etl army mil tecsun1 tec army mil wmich edu tecsun1 tec army mil on ion cmu edu tecsun1 tec army mil onion cmu edu cmu edu tecsun1 tec army mil d escartes etl army milwho s next mormons and jewscantaloupe srv cmu edu sci ta lk politics talk religion misc cantaloupe srv cmu edu rochester udel gatech h owland reston ans net usc utexas edu uunet iris mbvlab wpafb mil blackbird af it mil afterlife tecsun descartes riggs sci skeptic talk politics misc talk r eligion misc misc legal message apr gmt followup sci skeptic talk politics mi sc talk religion misc misc legal lnk corporation riverdale nntp posting desca rtes tec army mil was slightly surprised see the guns roston open here but ha s his posting record and have mine although usually more polite than not afra id controversy either this looks like argument that started just contrary pan der the sci skeptic masses will certainly bow out schedule for the next two w eeks does not read april through debate roston the true meaning the declarati on independence and the bill rights include long discourse the validity claim s the declaration independence have crossposted this misc legal that real law yers can pass judgement legal arguments article article one thing that should made clear that neither the fbi nor the batf responsible for what happened ye sterday one can argue about the initial raid but would worth mentioning befor e the facts get lost that the branch davidians were tipped off that the batf was coming during the initial raid the branch davidians opened fire first was waiting some some not intelligent person bring this look this country laws qu ote piece parchment that many seem think little amendment the right the peopl e secure their persons houses papers and effects against unreasonable searche s and seizures shall not violated and warrants shall issue but upon probable cause supported oath affirmation and particularly describing the place search ed and the persons things seized the first question did the notip off call in dicate the branch davidians that knock warrant was use and they could therefo re presumably into weapons free once the compound property was entered the se cond question what makes think that the davidians actions would have been dif ferent had another type warrant been use would like see both these questions answered before seeking judgment any claims about knock warrant clear violati on the amendment okay what about the fact that they were tipped off they shou ld not have opened fire right wrong think about drug dealer and competition w ants away with they call and tell that the feds are their way with knock warr ant being moronic sheep wait with our guns holstered now instead the feds com es competition and are history the only acceptable answer knock warrant blazi ng guns may sound paranoid but our government out control and killing few fed eral officers make knock some sense back into spurious analogy especially for one skeptical the davidians were not drug dealers and all accounts feared one but the feds any evidence the contrary when dangerous faction starts opinions not represent with wrath and vengeance their hearts those employer solemn lea gue and cov nant bound any government agency ruin slaughter and confound turn religion fable and turn the government babel the law disgrace the gown corrup t the senate rob the crown sacrifice old england glory and make her infamous story when such tempest shook the land how could unguarded virtue stand

1/1 [===================================
Actual Label - talk.politics.misc
Predicted Label - talk.religion.misc

unisql uucp unisql uucp unisql uucp srgenprp srgenprpwhere can i get a new yo rk taxicantaloupe srv cmu edu magnesium club cmu edu news sei cmu edu cis ohi o state edu pacific mps ohio state edu zaphod mps ohio state edu utexas edu u nisql wrat rec autos message apr gmt unisql inc austin texas usa article dork

sciences sdsu edu larc sdsu edu eagle lerc nasa gov eagle lerc nasa gov bach lerc nasa gov sciences sdsu edu ucssun1 sdsu eduelectric power line balls sci electronics message apr gmt article larc babb followup sci electronics usa sd su larc nntp posting larc sdsu edu article got question from dad which really can not answer and would appreciate some net wisdom his question about some d iameter balls which are attached electric power lines his area seen half doze n between two poles neither have any experience with electric power distribut ion only guess was that they may capacitive device equalize the inductance the grid but why many between two poles will bet there runway nearby sdsu larc

umbc edu umbc7 umbc edu umbc edulooking for an r5 xserver for hp9000 385canta loupe srv cmu edu rochester udel news udel edu darwin sura net haven umd edu news umbc edu nobody comp windows apr university maryland baltimore county ca mpus message nntp posting umbc umbc edu auth banz anyone know where can find one binaries are nice but source would thanks advance give people alternative microsoft will have been greater good yes are your father ibm presentation fo se

umd edu mimsy umd edu cmu edu network cyberia win net mimsy umd edu cyberia w in net umd educlinton wants national id card aka ussr style internal passport cantaloupe srv cmu edu talk politics talk politics alt fan rush talk politics guns talk politics drugs alt fan rush limbaugh message apr gmt followup talk politics guns the reality liberation front how bout embed the card the forhea d everyones skull not lose without being already dead close birth implant sma rt chip just behind the ear under the skin but above the skull incase hypo al lergenic high carbon content glass this chip would reprogrammed age with the pertinent medical correctional taxational data behave yourself will input int o your permenant record you forgot the part about encasing small shaped charg e that anyone tries tamper with explodes and kills you and the shaped charge can set off remote control but only you get out line properly patriotic citiz ens have nothing fear born die your laws off brain what funny bout peace love and understanding this sig contains animal products and was not tested animal s congress shall make law respecting establishment religion prohibiting the f ree exercise thereof abridging the freedom speech the press the right the peo ple peaceably assemble and petition the government for redress grievances the constitution the united states america

```
1/1 [=======] - 0s 20ms/step
```

aurora alaska edu aurora alaska edu news duc auburn edu raven alaska edu news duc auburn edu eng auburn edu eng auburn edu acad3 alaska edusolar sail datac antaloupe srv cmu edu crabapple srv cmu edu andrew cmu edu news sei cmu edu c is ohio state edu zaphod mps ohio state edu uwm edu ogicse news washington ed u raven alaska edu aurora alaska edu nsmca sci space message apr gmt article aurora apr university alaska fairbanks nntp posting acad alaska edu article l ooking for any information concerning projects involving solar sails understa nd that the jpl did extensive study the subject back the late but having trou ble gathering such information are there any groups out there currently invol ved such project know someone had long talks about solar sails early this year and late last year also about solar sailing not sure who captured possible think was one the regulars who had most all the data think started the latest round the late last year round but the topic has been around here off and for year two not high just jacked

ccc amdahl juts ccc amdahl byuvm bitnet juts ccc amdahl byuvm bitnet byuvm bi tnet byuvm bitnetrw is robert weiss the only orthodox christiancantaloupe srv cmu edu magnesium club cmu edu news sei cmu edu ece cmu edu europa eng gtefsd com howland reston ans net usc utexas edu uunet amdahl juts agr talk religion misc message apr gmt reply amdahl corporation sunnyvale article you keep maki ng references orthodox belief and saying things like held that what exact bod y theology are you drawing for what you call orthodox who that holds that mea nt what you said meant whenever your personal interpretation biblical passage s challenged your only response seems that one needs merely look the bible or der see the truth but what those who see biblical things differently from you are simply assume that you are the only one who really understands just curio us when refers the orthodox talking about the historical position the christi an faith such things are derived from biblcal texts through the centuries the apocolic fathers the faith you are right that people read things differently the bible and this alright parts like parables and such forth however when co mes the essential doctrines the historical orthodox christan beliefs there on ly one correct way read for example either the doctrine the trinity true fals e yes people read the texts differently but only one position true they both cannot according the text the doctrine true and has always existed therefore when people like come along with vision and thinks can undo centuries doctrin e that supported the bible people consider him cult

uiuc edu geneva rutgers edu geneva rutgers edu aramis rutgers edu uiuc educel l church discussion groupcantaloupe srv cmu edu crabapple srv cmu edu andrew cmu edu news sei cmu edu ece cmu edu europa eng gtefsd com howland reston ans net usc rutgers igor rutgers edu geneva rutgers edu christian soc religion ch ristian message apr gmt university illinois dept comp sci beginning mail disc ussion group about cell churches you are follower and are cell church church that transitioning cell church just interested learning more about cell churc hes send mail fool who gives what cannot keep gain what cannot lose

scicom alphacdc com scicom alphacdc com cmu edu cmu edu flb optiplan scicom a lphacdc comfirst spacewalkcantaloupe srv cmu edu rochester udel louie newsser ver jvnc net howland reston ans net usenet ins cwru edu magnus acs ohio state edu csn csn scicom wats sci space message apr gmt sci alpha science computer network denver article one time there was speculation that the first spacewal k was staged fake has any evidence support contradict this claim emerged was this claim perhaps another fevered cold war hallucination for one would avid reader sci space ussr what really happened item extract

sdc boeing fury boeing com news washington edu mojo intellection sdc boeing m ojo intellection intellection news washington edu stein washington edu sun co m sdc boeing sdc boeingproblem with r4 releasecantaloupe srv cmu edu comp win dows comp windows comp windows comp windows intrinsics message may gmt follow up comp windows boeing computer services seattle nntp posting zulu article ar ticle unidentified external symbol use lxaw lxext lxt not this order but prop er order wondering there has some changes link libraries should include some more libraries any help this matter will very useful and highly appreciated g et this too you put bstatic into your options goes away things seem run witho ut must not called very often from the sunos sun fixed shared library bug whi ch conflicts with the way builds the shared library causing these symbols not ably undefined when building some compiling bstatic bdynamic appears work sol ve the problem you are using openwindows please contact your local sun office and request the following patch description openwindows libxt jumbo patch ope nwindows undefined symbols when using shared lib source patch for use with th e mit libraries was developed retrofits into some fixes made get around this problem the patch export contrib sunos patch_version client server tech servi ces boeing computer services box seattle

zoo toronto edu zoo toronto edu aio jsc nasa gov tm0006 lerc nasa gov cbl umd edu cbl umd edu starburst umd edu zoo toronto eduspace station redesign jsc a lternative 4sci space cantaloupe srv cmu edu magnesium club cmu edu news sei cmu edu cis ohio state edu zaphod mps ohio state edu malgudi oar net news ans net europa eng gtefsd com howland reston ans net usc utexas edu utnut utzoo h enry message sat apr gmt toronto zoology article the only thing that scares t

he part about simply strapping ssme and nosecone and just launching have this vision something going terribly wrong with the launch resulting the complete loss the new modular space station does not make whole lot difference actuall y since they were not building spares the station hardware anyway least this only one launch fail svr resembles high speed collision toronto zoology betwe en svr and sunos utzoo henry

fsphy1 physics fsu edu world std mailer fsu edu fsphy1 physics fsu edu fsphy1 physics fsu edu world std world std fsphy1 physics fsu edumiscelaneous soon t o have baby questionssci med cantaloupe srv cmu edu crabapple srv cmu edu ece cmu edu europa eng gtefsd com howland reston ans net wupost sdd com saimiri p rimate wisc edu eng ufl edu usenet ufl edu mailer fsu edu fsphy physics fsu e du picl florida state university school higher thought wed apr gmt news vax v ms vnews message reply all article writes are about have our first baby and h ave few questions that dont seem able get answered our satisfaction reguardin g having baby boy circumsized what are the medical pros and cons all have hea rd its the parents unfortonately that truly about the best summation the rese arch that there advantages stated circumcison included probably prevention pe nile cancer simplicity personal hygiene prevention urinary tract infections a nd prevention unretractible foreskin disadvantages include infection from the procedure pain etc apologize trying pull this off the top head will post what discovered research did paper the topic research class nursing school really decision that the parents some parents use the reasoning that they will look like daddy and like their friends justification there nothing wrong with this just sure what you want since rather difficult uncircumcise male although maj or surgical procedure exists how about the pregnant woman sitting tub water h ave heard stories infection etc how about after the water has broken long you r membranes have not broken and you have not had any problems with your pregn ancy should sit tub water however would recommend using your own bathtub your own home nearly impossible guarantee the cleanliness and safety public hot tu bs nice warm bath can very relaxing especially your back killing you and woul d possibly advisable avoid bubble bath soap esp you are prone yeast infection hope these tips help you some

venus iucf indiana edu usenet ucs indiana edu usenet ucs indiana edu venus iu cf indiana edu vax5 cit cornell edu laura harvard edu stafford winona msus ed u stafford winona msus edu vax2 winona msus edu vax5 cit cornell edushaft dri ves and wheeliesrec motorcycles cantaloupe srv cmu edu magnesium club cmu edu news sei cmu edu cis ohio state edu pacific mps ohio state edu zaphod mps ohi o state edu sol ctr columbia edu usenet ucs indiana edu venus iucf indiana ed u brown message followup rec motorcycles news vax vms vnews nntp posting venu s iucf indiana edu reply iucf rec tue apr gmt article writes apr gmt possible wheelie motorcycle with shaft drive yes but the rear wheel comes off the grou nd not the front see just hops into the air figure you can wheelies with shaft drive bike had bmw that was wheelie monster course did not have the initial

power burst just twist into the air had pop the clutch also had replace front fork seals few times well the fairing bit heavy slamming down onto those litt le stantion tubes all the time but let give you fair trashed the ring pinion gear the final drive doing wheelies and this was cheap fix either there some kind slip device the shaft prevent from breaking unfortunately did not save t he gears the topic wheelies the other day saw kid big hurricane stoppy rear w heelie man had the rear end this bike about feet off the ground traffic light not recommend these activities anymore but looked damn impressive you can not keep both tires the ground least have pointed that direction cheers

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syl nec research nec research nec acs ucalgary myrddin imat myrddin imat seah unt imat acs ucalgary acs ucalgary syl necmaxima chain waxrec motorcycles nec systems laboratory inc fri apr gmt article article chain lubed wheel clean af ter careful was similarly impressed with first but recently have started find ing rust spots the chain went back least the sucker never rusted when was usi ng the bluegoo spray the chain wax onto the rollers and sideplates occassionally and rust will not problem later dod fxwg wide glide red lady baby bike now why would nec needs name agree with any this anyway was raised pack wild corn dogs

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