#### importing all libraries

```
In [1]: import os
        import re
        import sys
        import nltk
        import itertools
        import numpy as np
        import pandas as pd
        import seaborn as sns
        from sklearn import tree
        from sklearn.svm import SVC
        from joblib import dump, load
        from wordcloud import WordCloud
        import matplotlib.pyplot as plt
        from nltk.corpus import stopwords
        from nltk.stem import PorterStemmer
        from nltk.stem import WordNetLemmatizer
        from keras.models import Sequential,Model
        from sklearn.metrics import accuracy score
        from sklearn.metrics import confusion_matrix
        from sklearn.naive_bayes import MultinomialNB
        from sklearn.preprocessing import LabelEncoder
        from keras.utils.np utils import to categorical
        from sklearn.ensemble import AdaBoostClassifier
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.model selection import train test split
        from sklearn.feature extraction.text import CountVectorizer
        from keras.layers import Dense, LSTM, Spatial Dropout1D, Embedding
```

#### **Reading the Dataset**

Out[2]:

In [2]: df = pd.read\_csv('/kaggle/input/resume-dataset/UpdatedResumeDataSet.csv')
df

	Category	Resume
0	Data Science	Skills * Programming Languages: Python (pandas
1	Data Science	Education Details \r\nMay 2013 to May 2017 B.E
2	Data Science	Areas of Interest Deep Learning, Control Syste
3	Data Science	Skills • R • Python • SAP HANA • Table
4	Data Science	Education Details \r\n MCA YMCAUST, Faridab
957	Testing	Computer Skills: • Proficient in MS office (
958	Testing	â Willingness to accept the challenges. â
959	Testing	PERSONAL SKILLS • Quick learner, • Eagerne
960	Testing	COMPUTER SKILLS & SOFTWARE KNOWLEDGE MS-Power
961	Testing	Skill Set OS Windows XP/7/8/8.1/10 Database MY

962 rows × 2 columns

### **List of all Categories**

```
for i in range(len(df['Category'].unique())):
    print(df['Category'].unique()[i])
Data Science
Advocate
Arts
Web Designing
Mechanical Engineer
Sales
Health and fitness
Civil Engineer
Java Developer
Business Analyst
SAP Developer
Automation Testing
Electrical Engineering
Operations Manager
Python Developer
DevOps Engineer
Network Security Engineer
PMO
Database
Hadoop
ETL Developer
DotNet Developer
Blockchain
Testing
```

# Visualizing the data distribution in each category.

```
In [4]: plt.figure(figsize=(50, 20), dpi=130)
for i in range(len(df['Category'].unique())):
    plt.bar(df['Category'].unique()[i], len(df['Category']==df['Category'].
plt.show()
```

# Visualizing most commonly used words in each type of Resumes

#### Words Commonly Used in DataScience Resumes



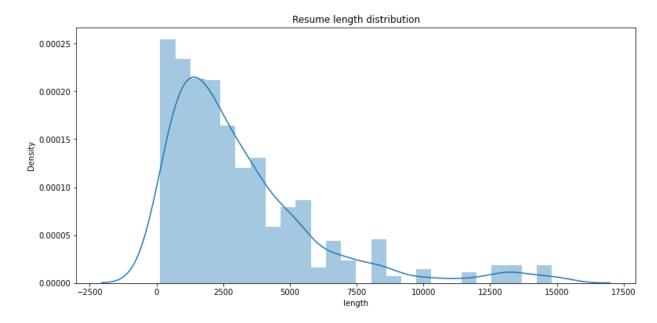
Visualizing the length disstribution of the content in the Resumes.

```
In [6]: df['length'] = df['Resume'].str.len()
    plt.figure(figsize=(12.8,6))
    sns.distplot(df['length']).set_title('Resume length distribution')
```

/opt/conda/lib/python3.7/site-packages/seaborn/distributions.py:2619: Fut ureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[6]: Text(0.5, 1.0, 'Resume length distribution')

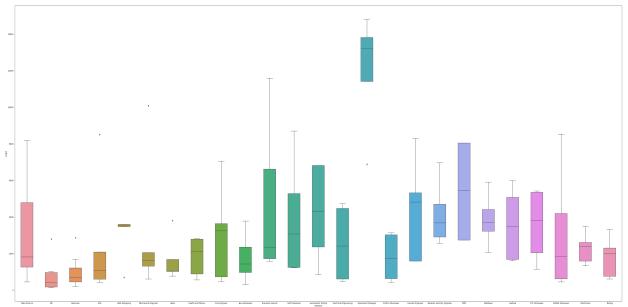


### **Summary of Length Distribution**

```
In [7]: df['length'].describe()
Out[7]: count
                    962.000000
                   3160.364865
        mean
        std
                   2886.528521
                    142.000000
        min
        25%
                   1217.250000
        50%
                   2355.000000
        75%
                   4073.750000
                  14816.000000
        max
        Name: length, dtype: float64
```

### Visualizing of number of words in each category of resume using boxplot

```
In [8]: plt.figure(figsize=(60,30))
sns.boxplot(data=df, x='Category', y='length', width=.5);
```



### **Pre Processing**

### **Checking for missing data**

```
In [9]: print(df.isnull().sum())

Category 0
Resume 0
length 0
dtype: int64
```

### Converting the data into lower case and removing words with small lengths

/opt/conda/lib/python3.7/site-packages/ipykernel\_launcher.py:7: SettingWi
thCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy)
import sys

#### removing punctuations

/opt/conda/lib/python3.7/site-packages/ipykernel\_launcher.py:5: FutureWar ning: The default value of regex will change from True to False in a futu re version. In addition, single character regular expressions will \*not\* be treated as literal strings when regex=True.

# Removing '\n' and '\t', extra spaces, quoting text and progressive pronouns

#### **Applying Lemmatization**

```
In [13]: |nltk.download('punkt')
         nltk.download('wordnet')
         nltk.download('omw-1.4')
         [nltk data] Downloading package punkt to /usr/share/nltk data...
         [nltk data]
                       Package punkt is already up-to-date!
         [nltk data] Downloading package wordnet to /usr/share/nltk data...
                       Package wordnet is already up-to-date!
         [nltk data]
         [nltk data] Downloading package omw-1.4 to /usr/share/nltk data...
Out[13]: True
In [14]: |wl = WordNetLemmatizer()
         nr = len(df)
         lis = []
         for r in range(0, nr):
             11 = []
             t = df.loc[r]['Resume']
             tw = str(t).split("")
             for w in tw:
                 11.append(wl.lemmatize(w, pos="v"))
             lt = " ".join(11)
             lis.append(lt)
In [15]: |df['Resume'] = lis
```

#### Removing Stop-words

/opt/conda/lib/python3.7/site-packages/ipykernel\_launcher.py:4: FutureWar ning: The default value of regex will change from True to False in a futu re version.

after removing the cwd from sys.path.

### Visualizing most commonly used words in Resumes after applying NLP techniques

#### Words Commonly Used in DataScience Resumes



#### **Data Preparation for Training and Testing**

#### **Encoding Labels**

```
In [19]: df['Category'] = LabelEncoder().fit_transform(df['Category'])
```

### Using Bag of Words approach for converting the content in resumes into vector form

```
In [20]: cv = CountVectorizer(max_features = 2500)
X = cv.fit_transform(df['Resume']).toarray()
y = df['Category']
```

#### Splitting the Data using Stratified split

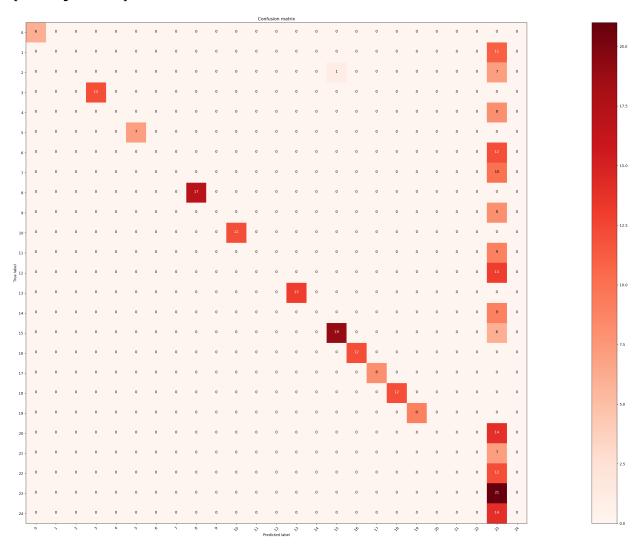
```
In [21]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3,
In [22]: def plot_confusion_matrix(cm, classes,
                                    normalize=False,
                                   title='Confusion matrix',
                                        cmap=plt.cm.Greens):
             plt.figure(figsize=(50, 20), dpi=130)
             plt.imshow(cm, interpolation='nearest', cmap=cmap)
             plt.title(title)
             plt.colorbar()
             tick_marks = np.arange(len(classes))
             plt.xticks(tick_marks, classes, rotation=45)
             plt.yticks(tick_marks, classes)
             if normalize:
                 cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
             thresh = cm.max() / 2.
             for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
                 plt.text(j, i, cm[i, j],
                          horizontalalignment="center",
                          color="white" if cm[i, j] > thresh else "black")
             plt.tight_layout()
             plt.ylabel('True label')
             plt.xlabel('Predicted label')
```

# Using AdaBoost Classifier as the Model and printing evaluating it using confusion matrix

```
In [23]: clf = AdaBoostClassifier(n_estimators=60)
    clf = clf.fit(X_train, y_train)
    yp = clf.predict(X_test)
    acc = accuracy_score(y_test, yp)
    print("accuracy is: ",acc)
    CM = confusion_matrix(y_test, yp)
    plot_confusion_matrix(CM, classes = range(25),cmap=plt.cm.Reds)
    dump(clf, 'ada.joblib')
```

accuracy is: 0.5121107266435986

#### Out[23]: ['ada.joblib']

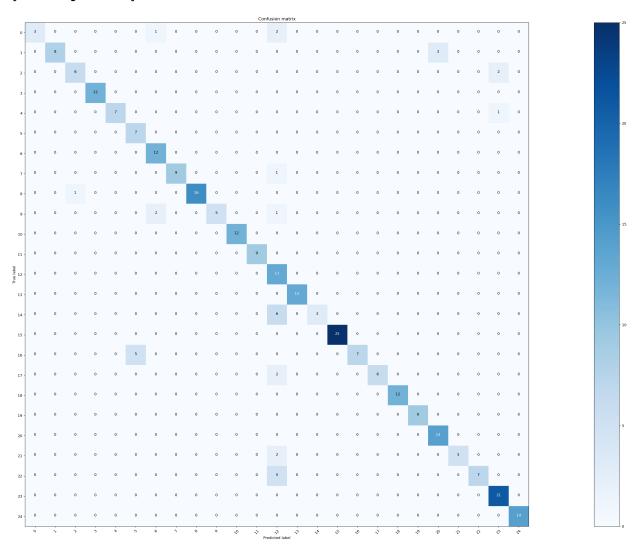


# Using KNeighbors Classifier as the Model and printing evaluating it using confusion matrix

```
In [24]: clf = KNeighborsClassifier(n_neighbors=7)
    clf = clf.fit(X_train, y_train)
    yp = clf.predict(X_test)
    acc = accuracy_score(y_test, yp)
    print("accuracy is: ",acc)
    CM = confusion_matrix(y_test, yp)
    plot_confusion_matrix(CM, classes = range(25),cmap=plt.cm.Blues)
    dump(clf, 'knei.joblib')
```

accuracy is: 0.8823529411764706

```
Out[24]: ['knei.joblib']
```

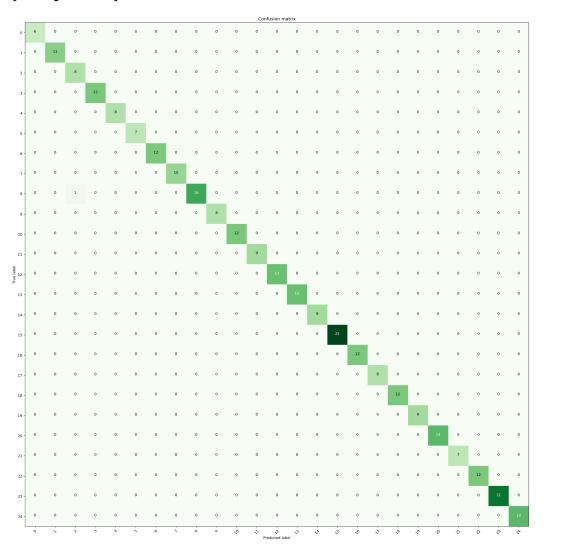


# Using Decision tree as the Model and printing evaluating it using confusion matrix

```
In [25]: clf = tree.DecisionTreeClassifier()
    clf = clf.fit(X_train, y_train)
    yp = clf.predict(X_test)
    acc = accuracy_score(y_test, yp)
    print("accuracy is: ",acc)
    CM = confusion_matrix(y_test, yp)
    plot_confusion_matrix(CM, classes = range(25))
    dump(clf, 'DT.joblib')
```

accuracy is: 0.9965397923875432

Out[25]: ['DT.joblib']



### **Turning encoded lables into Categorical** format

```
In [26]: y_train=to_categorical(y_train, num_classes = 25, dtype='float32')
    y_test=to_categorical(y_test, num_classes = 25, dtype='float32')

In [27]: from keras.models import Sequential,Model
    from tensorflow.keras.utils import plot_model
    from keras.layers import Dense,LSTM, SpatialDropout1D, Embedding
    from keras.layers import Dense, Embedding, GRU, LSTM, Dropout, Bidirectiona
    from keras.layers import Conv1D, MaxPool1D, GlobalMaxPooling1D, GlobalAvera
```

### **Building a 1D CNN model**

```
In [28]: model = Sequential()
    model.add(Embedding(input_dim=232337, output_dim=100, input_length=X_train.
    model.add(Conv1D(128, 3, activation='relu'))
    model.add(Dropout(0.2))
    model.add(Conv1D(128, 3, activation='relu'))
    model.add(GlobalMaxPooling1D())
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
    model.add(Dropout(0.2))
```

2022-10-07 21:50:28.136797: I tensorflow/stream\_executor/cuda/cuda\_gpu\_ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:28.235594: I tensorflow/stream\_executor/cuda/cuda\_gpu\_ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:28.236381: I tensorflow/stream\_executor/cuda/cuda\_gpu\_ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:28.237985: I tensorflow/core/platform/cpu\_feature\_guard. cc:142] This TensorFlow binary is optimized with oneAPI Deep Neural Netwo rk Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 AVX512F FMA

To enable them in other operations, rebuild TensorFlow with the appropria te compiler flags.

2022-10-07 21:50:28.238320: I tensorflow/stream\_executor/cuda/cuda\_gpu\_ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:28.239149: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:28.239849: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:30.567415: I tensorflow/stream executor/cuda/cuda qpu ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:30.568344: I tensorflow/stream\_executor/cuda/cuda\_gpu\_ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:30.569033: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-10-07 21:50:30.569631: I tensorflow/core/common runtime/gpu/gpu devi ce.cc:1510] Created device /job:localhost/replica:0/task:0/device:GPU:0 w ith 15401 MB memory: -> device: 0, name: Tesla P100-PCIE-16GB, pci bus i d: 0000:00:04.0, compute capability: 6.0

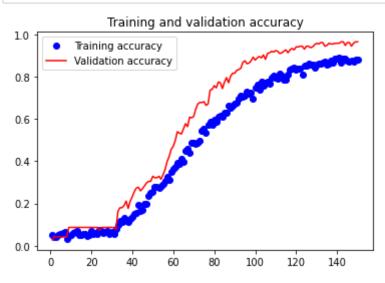
In [29]: history = model.fit(X\_train, y\_train, batch\_size = 64, epochs=150, validati

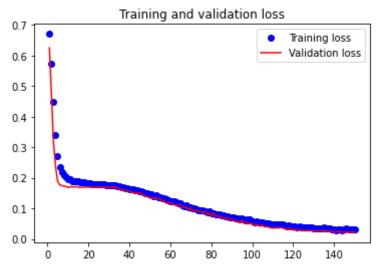
2022-10-07 21:50:31.455512: I tensorflow/compiler/mlir\_graph\_optimiz ation\_pass.cc:185] None of the MLIR Optimization Passes are enabled (registered 2)

Epoch 1/150

2022-10-07 21:50:33.007257: I tensorflow/stream\_executor/cuda/cuda\_dnn.c c:369] Loaded cuDNN version 8005

```
acc = history.history['accuracy']
In [30]:
         val acc = history.history['val accuracy']
         loss = history.history['loss']
         val loss = history.history['val loss']
         epochs = range(1, len(acc) + 1)
         plt.plot(epochs, acc, 'bo', label='Training accuracy')
         plt.plot(epochs, val_acc, 'r', label='Validation accuracy')
         plt.title('Training and validation accuracy')
         plt.legend()
         plt.figure()
         plt.plot(epochs, loss, 'bo', label='Training loss')
         plt.plot(epochs, val_loss, 'r', label='Validation loss')
         plt.title('Training and validation loss')
         plt.legend()
         plt.show()
```





In [32]: model.save('cnn.h5')