```
In [1]: import warnings
        warnings.filterwarnings('ignore')
In [2]: import pandas as pd
        import numpy as np
         import re
         import nltk
         import matplotlib.pyplot as plt
         import seaborn as sns
         %matplotlib inline
         from nltk.corpus import stopwords
        from nltk.stem.porter import PorterStemmer
         from sklearn.svm import SVC
         from sklearn.preprocessing import LabelEncoder
         from sklearn.metrics import confusion matrix
In [3]: df = pd.read csv('Resume dataset.csv')
        df.head()
In [4]:
Out[4]:
              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 â□¢ R â□¢ Python â□¢ SAP HANA â□¢ Table...
         4 Data Science
                           Education Details \r\n MCA YMCAUST, Faridab...
In [5]: df.isnull().sum()
Out[5]: Category
         Resume
         dtype: int64
In [6]: print("Displaying the distinct categories of resume -\n")
         print(df['Category'].unique())
       Displaying the distinct categories of resume -
       ['Data Science' 'HR' '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']
In [7]: print("Displaying the number of records belonging to distinct categories of resume
         print(df['Category'].value_counts())
```

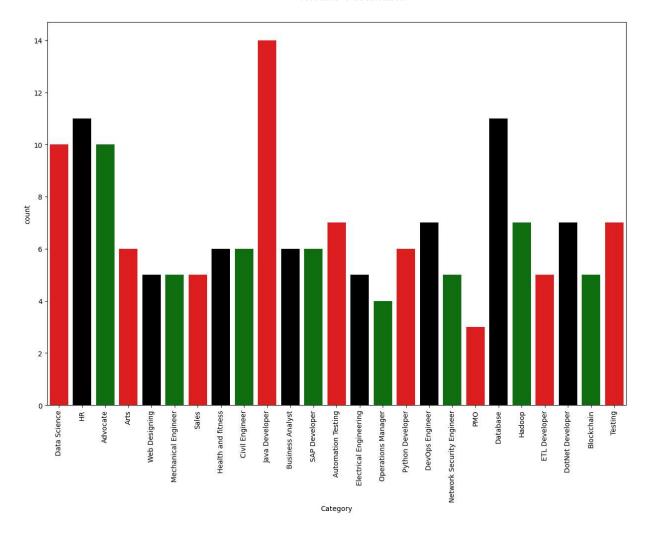
Displaying the number of records belonging to distinct categories of resume -

```
Category
       Java Developer
                                     14
       Database
                                     11
       HR
                                     11
       Data Science
                                     10
       Advocate
                                     10
       DotNet Developer
                                      7
       Hadoop
                                      7
       DevOps Engineer
                                      7
                                      7
       Automation Testing
       Testing
                                      7
       Civil Engineer
                                      6
       Business Analyst
                                      6
       SAP Developer
                                      6
       Health and fitness
                                      6
       Python Developer
                                      6
       Arts
                                      6
       Electrical Engineering
                                      5
                                      5
                                      5
       Network Security Engineer
                                      5
       Mechanical Engineer
                                      5
       Web Designing
       ETL Developer
                                      5
       Blockchain
                                      5
       Operations Manager
                                      4
       PMO
                                      3
       Name: count, dtype: int64
In [8]: df['Category'].nunique()
        Visualization of category
```

Out[8]: 25

```
In [9]: plt.figure(figsize = (15, 10))
        plt.xticks(rotation = 90)
        sns.countplot(x = 'Category', data = df, palette = ['red', 'black', 'green'])
```

Out[9]: <Axes: xlabel='Category', ylabel='count'>



Label Encoding

```
In [10]: le = LabelEncoder()
    df['Category'] = le.fit_transform(df['Category'])
    df.head()
```

Out[10]:		Category	Resume
	0	6	Skills * Programming Languages: Python (pandas
	1	6	Education Details \r\nMay 2013 to May 2017 B.E
	2	6	Areas of Interest Deep Learning, Control Syste
	3	6	Skills â□¢ R â□¢ Python â□¢ SAP HANA â□¢ Table
	4	6	Education Details \r\n MCA YMCAUST, Faridab

Text Preprocessing

```
In [11]: ps = PorterStemmer()
    stopwords = nltk.corpus.stopwords.words('english')
    def get_clean(x):
        x = re.sub('[^a-zA-Z]', ' ', x).lower().split()
```

```
Resume Classification
              x = ' '.join([ps.stem(word) for word in x if word not in stopwords])
              return x
In [12]:
          df['Resume'] = df['Resume'].apply(lambda x: get clean(x))
In [13]:
         df.head()
Out[13]:
             Category
                                                           Resume
          0
                     6 skill program languag python panda numpi scipi...
          1
                     6
                          educ detail may may b e uit rgpvdata scientist...
          2
                     6
                          area interest deep learn control system design...
          3
                     6
                          skill r python sap hana tableau sap hana sql s...
                     6 educ detail mca ymcaust faridabad haryanadata ...
          4
          TF-IDF Vectorizer
In [14]: from sklearn.feature extraction.text import TfidfVectorizer
          tfidf = TfidfVectorizer(max features = 3000, ngram range = (1, 3))
          X = tfidf.fit transform(df['Resume'])
          y = df['Category']
In [15]: print(tfidf.get_feature_names_out())
        ['abap' 'abil' 'abil work' ... 'year work' 'york' 'zone']
In [16]: data sample 1 = df[:2]
          tfidf1 = TfidfVectorizer()
          data_2_tfidf = tfidf1.fit_transform(data_sample_1['Resume'])
In [17]: data_2_tfidf.shape
Out[17]: (2, 346)
In [18]: pip install --upgrade scikit-learn
```

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: scikit-learn in c:\users\ds_lab_18\appdata\roaming\py thon\python312\site-packages (1.5.1)

Requirement already satisfied: numpy>=1.19.5 in c:\programdata\anaconda3\lib\site-pa ckages (from scikit-learn) (1.26.4)

Requirement already satisfied: scipy>=1.6.0 in c:\programdata\anaconda3\lib\site-pac kages (from scikit-learn) (1.13.1)

Requirement already satisfied: joblib>=1.2.0 in c:\programdata\anaconda3\lib\site-pa ckages (from scikit-learn) (1.4.2)

Requirement already satisfied: threadpoolctl>=3.1.0 in c:\users\ds_lab_18\appdata\ro aming\python\python312\site-packages (from scikit-learn) (3.5.0)

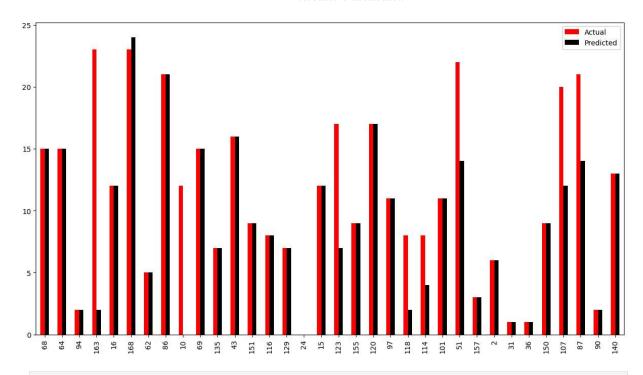
Note: you may need to restart the kernel to use updated packages.

```
[notice] A new release of pip is available: 24.1.2 -> 24.2
        [notice] To update, run: python.exe -m pip install --upgrade pip
In [19]: df_tfidf = pd.DataFrame(data_2_tfidf.toarray(), columns = tfidf1.get_feature_names_
         df tfidf.head()
Out[19]:
              acceler account
                                 achiev
                                                   action
                                                           address
                                                                      aditya
                                          across
                                                                              advanc
                                                                                          also
         0 0.026709 0.026709 0.000000 0.080128 0.026709
                                                          0.026709 0.000000 0.026709 0.026709
          1 0.000000 0.000000 0.069665 0.000000 0.000000 0.000000 0.069665 0.000000 0.000000
         2 rows × 346 columns
In [20]: from sklearn.model_selection import train_test_split
         from sklearn.metrics import accuracy_score
         X train, X val, y train, y val = train test split(X, y, test size = 0.20)
In [21]: print('X_train shape :', X_train.shape)
         print('y_train shape :', y_train.shape)
         print('X val shape :', X val.shape)
         print('y_val shape :', y_val.shape)
        X_train shape : (135, 3000)
        y_train shape : (135,)
        X_val shape : (34, 3000)
        y val shape : (34,)
         Logistic Regression
In [22]: from sklearn.linear model import LogisticRegression
         log = LogisticRegression()
         log.fit(X_train, y_train)
         y_pred_log = log.predict(X_val)
In [23]: print('Accuracy of Logistic Classifier: {:.2f}'.format(accuracy_score(y_val, y_pred
        Accuracy of Logistic Classifier: 0.41
         Grid Search
In [24]: from sklearn.model_selection import GridSearchCV
In [25]: log = LogisticRegression()
         grid_values = {'penalty':['l1', 'l2'], 'C':[0.001, 0.01, 1, 5, 10, 25]}
         grid_log_acc = GridSearchCV(log, param_grid = grid_values)
         grid_log_acc.fit(X_train, y_train)
```

```
Out[25]: | >
                                                                                        GridSearchCV
                                          ▶ best_estimator_: LogisticRegression
                                                                             LogisticRegression
In [26]: #Predict values based on new parameters
                                    y pred log acc = grid log acc.predict(X val)
                                     #New Model Evaluation Metrics
                                     print('Accuracy Score: ', accuracy_score(y_val, y_pred_log_acc))
                                Accuracy Score: 0.7647058823529411
In [27]: df log = pd.DataFrame({'Actual': y val, 'Predicted': y pred log acc})
In [28]: df_log.plot(kind = 'bar', figsize = (15, 8), color = ['lightcoral', 'skyblue'])
                                      plt.show()
                                                                                                                                                                                                                                                                                                                                     Predicted
                               20
                               10
                                                                    163 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 168 - 
                                                                                                                                                                                                                                                                                                                    150
                                    SVC
In [29]: classifier = SVC()
                                     classifier.fit(X_train, y_train)
Out[29]:
                                     ▼ SVC 1 ?
                                    SVC()
In [30]: y_pred_svc = classifier.predict(X_val)
```

```
In [31]: print('Accuracy of SVC Classifier: {:.2f}'.format(accuracy_score(y_val, y_pred_svc))
        Accuracy of SVC Classifier: 0.29
         Grid Search
         param_grid = {'C':[0.1,1, 10, 100, 1000], 'gamma':[1, 0.1, 0.01, 0.001, 0.0001]}
In [32]:
In [33]: grid svc acc = GridSearchCV(classifier, param grid)
         grid svc acc.fit(X train, y train)
         #Predict values based on new parameters
         y_pred_svc_acc = grid_svc_acc.predict(X_val)
         #New Model EValuation metrics
         print('Accuracy Score : ', accuracy_score(y_val, y_pred_svc_acc))
        Accuracy Score : 0.6470588235294118
In [34]: df_svc = pd.DataFrame({'Actual': y_val, 'Predicted': y_pred_svc_acc})
In [35]: df svc.plot(kind = 'bar', figsize = (15,8), color = ['purple', 'yellow'])
         plt.show()
                                                                                        Actual
                                                                                        Predicted
        20
        15
        10
         Random Forest
In [36]: from sklearn.ensemble import RandomForestClassifier
In [37]: clf rf = RandomForestClassifier(n estimators = 200)
         clf rf.fit(X train, y train)
```

```
Out[37]:
                  RandomForestClassifier
         RandomForestClassifier(n estimators=200)
In [38]: y_pred_rf = clf_rf.predict(X_val)
In [39]: print('Accuracy of SVC Classifier : {:.2f}'.format(accuracy_score(y_val, y_pred_rf)
        Accuracy of SVC Classifier: 0.74
         Grid Search
In [40]: param_grid_rf = {"n_estimators": np.arange(2, 300, 2)}
In [41]: from sklearn.model selection import GridSearchCV
In [48]: grid_rf_acc = GridSearchCV(clf_rf, param_grid_rf)
In [49]: grid_rf_acc.fit(X_train, y_train)
Out[49]: | •
                         GridSearchCV
          ▶ best_estimator_: RandomForestClassifier
                  RandomForestClassifier
In [50]: #predict values based on new parameters
         y_pred_rf_acc = grid_rf_acc.predict(X_val)
         #New Model Evaluation Metrics
         print('Accuracy Score : ', accuracy_score(y_val, y_pred_rf_acc))
        Accuracy Score: 0.7352941176470589
In [51]: df rf = pd.DataFrame({'Actual': y val, 'Predicted': y pred rf acc})
In [58]: | df_rf.plot(kind = 'bar', figsize = (15, 8), color = ['red', 'black'])
         plt.show()
```



```
In [53]: y_val_inverse = le.inverse_transform(y_val)
```

In [54]: predicted_inverse = le.inverse_transform(y_pred_log_acc)

Out[55]:		Actual	Predicted
	1	Java Developer	Java Developer
	2	Automation Testing	Automation Testing
	3	Testing	Java Developer
	4	HR	HR
	5	Testing	Automation Testing
	6	Civil Engineer	Civil Engineer

Single Prediction

```
In [60]: text = "KEY SKILLS: â€Â¢ Computerized accounting with tally â€Â¢ Sincere & hard

In [61]: text = get_clean(text)
    print(text)
```

key skill computer account talli sincer hard work manag account incom tax good commu n leadership two four wheeler drive licens internet ecommerc manag comput skill c la nguag web program talli dbm educ detail june june mba financ hr india mlrit june jun e bcom comput hyderabad telangana osmania univers june april inter mec india srimedh avhrnaniskil detail account exprienc month databas manag system exprienc month dbm e xprienc month manag account exprienc month ecommerc exprienc monthscompani detail co mpani valuelab descript give rrf form requir dlt hand rlt scrum master take form rlt scrum master give form traine work requir till candid receiv offer compani

```
In [62]: vec = tfidf.transform([text])
In [63]: predict = grid_log_acc.predict(vec)
         predict
Out[63]: array([12])
In [66]: y pred = le.inverse transform(predict)
         y_pred
Out[66]: array(['HR'], dtype=object)
         Another Example
In [67]: text ds = "Machine learning, Deep learning, Python, Statistics"
In [68]: text_ds = get_clean(text_ds)
         print(text ds)
        machin learn deep learn python statist
In [69]: vec = tfidf.transform([text_ds])
In [70]: predict = grid_log_acc.predict(vec)
         predict
Out[70]: array([6])
In [71]: y pred = le.inverse transform(predict)
         y_pred
Out[71]: array(['Data Science'], dtype=object)
In [ ]:
```