```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix
```

Connect to Google Drive

```
!rm -rf /root/.config/Google/DriveFS
!rm -rf /root/.config/Google/Drive/

from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

File_1 = pd.read_csv('/content/drive/My Drive/Resume data set/clean_resume_data.csv')
```

Import Clean DataSet

File_1.head()

→		ID	Category	Feature
	0	16852973	HR	hr administrator marketing associate hr admini
	1	22323967	HR	hr specialist hr operations summary media prof
	2	33176873	HR	hr director summary years experience recruitin
	3	27018550	HR	hr specialist summary dedicated driven dynamic
	4	17812897	HR	hr manager skill highlights hr skills hr depar

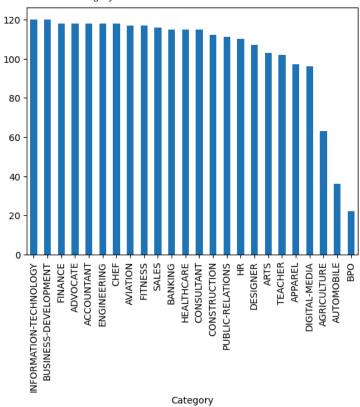
File_1.shape

→ (2484, 3)

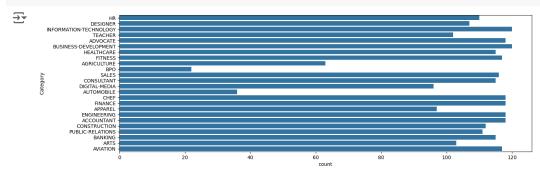
Exploratory Data Analysis

File_1['Category'].value_counts().plot(kind='bar')





```
plt.figure(figsize=(15,5))
sns.countplot(File_1['Category'])
plt.show()
```



```
counts = File_1['Category'].value_counts()
labels = File_1['Category'].unique()
plt.figure(figsize=(15,10))
plt.pie(counts,labels=labels,autopct='\%1.1f\%',shadow=True, colors=plt.cm.plasma(np.linspace(0,1,3)))
plt.show()
```

Show hidden output

Balance DataSet

```
from sklearn.utils import resample
# Define the maximum count among all categories
max_count = File_1['Category'].value_counts().max()
# Resample each category to match the maximum count
balanced data = []
for category in File_1['Category'].unique():
    category_data = File_1[File_1['Category'] == category]
    if len(category_data) < max_count:</pre>
        # Perform oversampling for categories with fewer samples
        balanced_category_data = resample(category_data, replace=True, n_samples=max_count, random_state=42)
    else:
        # Perform undersampling for categories with more samples
        balanced_category_data = resample(category_data, replace=False, n_samples=max_count, random_state=42)
    balanced_data.append(balanced_category_data)
# Concatenate the balanced data for all categories
balanced_df = pd.concat(balanced_data)
balanced_df['Category'].value_counts()
→ Category
                              120
     DESIGNER
                              120
     ARTS
                              120
     BANKING
                              120
     PUBLIC-RELATIONS
                              120
     CONSTRUCTION
                              120
     ACCOUNTANT
                              120
     ENGINEERING
                              120
     APPAREL
                              120
     FINANCE
                              120
     CHFF
                              120
     AUTOMOBILE
                              120
     DIGITAL-MEDIA
                              120
     CONSULTANT
                              120
     SALES
                              120
     BP0
                              120
     AGRICULTURE
                              120
     FITNESS
                              120
     HEALTHCARE
                              120
     BUSINESS-DEVELOPMENT
                              120
     ADVOCATE
                              120
     TEACHER
                              120
     INFORMATION-TECHNOLOGY
                              120
     AVIATION
                              120
     Name: count, dtype: int64
balanced df.isnull().sum()
   ID
    Category
                0
     Feature
                1
     dtype: int64
```

balanced_df.dropna(inplace=True)

Train & Test data

```
X = balanced_df['Feature']
y = balanced_df['Category']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Encoding (TF-IDF)

X_train

```
director information technology professional p...
senior finance manager summary highly driven f...
consultant summary media planner media buyer r...
adult education teacher accomplishments cal po...
banking summary high energy manager successful...
```

```
979 manager administration facilities executive su...

1027 pmc psm sales manager sales associate career f...

793 fitness specialist summary energetic personal ...

Name: Feature, Length: 2303, dtype: object

tfidf_vectorizer = TfidfVectorizer()

X_train_tfidf = tfidf_vectorizer.fit_transform(X_train)
```

Train Random Forest Classifier

X_test_tfidf = tfidf_vectorizer.transform(X_test)

director executive profile forward thinker exp...

bpo team leader summary seeking provide top qu...

1328

973

```
from sklearn.metrics import classification_report
rf_classifier = RandomForestClassifier()
rf_classifier.fit(X_train_tfidf, y_train)
# Step 4: Accuracy Evaluation
y_pred = rf_classifier.predict(X_test_tfidf)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
print(classification_report(y_test,y_pred))
# Confusion Matrix
conf_matrix = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(10, 8))
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues', xticklabels=rf_classifier.classes_, yticklabels=rf_classifier.classes_)
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix')
plt.show()
```

_				
→ 🕶	Accuracy:	0.	.84895833333	2

Accuracy: 0.8489583333333	3334 precision	recall	f1-score	support			
ACCOUNTANT	0.78	1.00	0.88	21			
ADVOCATE	0.91	0.69	0.78	29			
AGRICULTURE	0.95	0.78	0.86	23			
APPAREL	0.90	0.86 0.59	0.88	21			
ARTS	1.00		0.74	22			
AUTOMOBILE	1.00	1.00	1.00	19 30			
AVIATION BANKING	0.91 0.94	1.00 0.74 1.00 0.61	0.95 0.83 1.00 0.56	23			
BPO	1.00			15			
BUSINESS-DEVELOPMENT	0.52			18			
CHEF	0.90	0.96	0.93	28			
CONSTRUCTION	0.89	0.96	0.92	25			
CONSULTANT	1.00	0.68	0.81	31			
DESIGNER DIGITAL-MEDIA	0.96 0.83	0.93 0.90	0.95 0.86	28 21			
ENGINEERING	0.92	0.96	0.94	23			
FINANCE	0.76	0.62	0.68	21			
FITNESS	0.78	0.95	0.86	22			
HEALTHCARE	0.81	0.72	0.76	29			
HR	0.66	1.00	0.79	21			
INFORMATION-TECHNOLOGY	0.73	0.83	0.78	23			
PUBLIC-RELATIONS SALES	0.84 0.86	0.91 0.70	0.87 0.78	23 27			
TEACHER	0.79	1.00	0.88	33			
accuracy			0.85	576			
macro avg	0.86	0.85	0.85	576			
weighted avg	0.86	0.85	0.85	576			
			Confus	ion Matrix			
ACCOUNTANT -	21 0 0 0	0 0 0 0			0 0 0 0 0	0 0 0 0	
ADVOCATE -	0 20 0 0	0 0 0 0	0 0 2 0	0 0 0 0 1	0 0 1 3 0	0 0 1 1	
AGRICULTURE -	0 0 18 0	0 0 0 0	0 0 0 0	0 0 0 0 1	0 0 1 1 1	1 0 0 1	- 30
APPAREL -	0 0 0 18	0 0 0 0	0 0 0 0	0 0 0 0	0 1 0 1 0	0 0 1 0	
ARTS -	0 0 0 0	13 0 1 0	0 0 1 0	0 0 0 1 0	0 1 0 0 2	2 0 0 3	
AUTOMOBILE -	0 0 0 0	0 19 0 0	0 0 0 0	0 0 0 0	0 0 0 0 0	0 0 0 0	- 25
AVIATION -	0 0 0 0	0 0 30 0		0 0 0 0	0 0 0 0 0	0 0 0 0	
BANKING -	1 0 0 0	0 0 0 17	0 1 0 0	0 0 0 0	3 1 0 0 0	0 0 0 0	
BPO -	0 0 0 0	0 0 0 0	15 0 0 0	0 0 0 0	0 0 0 0 0	0 0 0 0	
BUSINESS-DEVELOPMENT -	0 0 0 0	0 0 0 0			0 0 0 1 2		- 20
CHEF -		0 0 0 0			0 0 0 0 0		
CONSTRUCTION -		0 0 0 0			0 0 1 0 0		
		0 0 0 0			1 1 0 1 1		- 15
DESIGNER -		0 0 1 0			0 1 0 0 0		
DIGITAL-MEDIA -		0 0 0 0		0 0 0 19 0	0 0 0 0 0		
ENGINEERING -	0 0 0 0	0 0 0 0	0 1 0 0	0 0 0 22	0 0 0 0	0 0 0	
FINANCE -		0 0 0 0	0 0 0 0		13 0 0 2 0		- 10
FITNESS -	0 0 0 0	0 0 0 0		0 0 0 0	0 21 0 0 0		
HEALTHCARE -		0 0 0 0	0 2 0 0		0 1 21 1 1		
HR - INFORMATION-TECHNOLOGY -		0 0 0 0			0 0 0 21 0		- 5
		0 0 0 0	0 0 0 0		0 0 2 1 1		
PUBLIC-RELATIONS -		0 0 1 1	0 0 0 0				
SALES - TEACHER -		0 0 0 0	0 2 0 2	0 0 0 0 0	0 0 0 0 0		
	1 1 1 1	1 1 1 1	1 1 1 1	1 1 1 1	1 1 1 1 1	1 1 1	- 0
	ANT ATE URE	ARTS DBILE ATION KING	BPO MENT CHEF	ANT NER IDIA	ESS ARE HR	SALES	
	ACCOUNTANT ADVOCATE AGRICULTURE APPAREL	ARTS AUTOMOBILE AVIATION BANKING	BPO SINESS-DEVELOPMENT CHEF	CONSULTANT DESIGNER DIGITAL-MEDIA ENGINEERING	FINANCE FITNESS HEALTHCARE HR	PUBLIC-RELATIONS SALES TEACHER	
	AD AB NRIC	OTO A	ÆLC	ONS DE ITAL	EAU CHN		
	A A	∢	-DE	E BIG	- E	ZIC Z	
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			SIN		M	L. C.	

Predicted

```
#Clean the resume
import re
def cleanResume(txt):
   cleanText = re.sub('http\S+\s', ' ', txt)
   cleanText = re.sub('RT|cc', ' ', cleanText)
cleanText = re.sub('#\S+\s', ' ', cleanText)
   cleanText = re.sub('@\S+', ' ', cleanText)
   cleanText = re.sub('\s+', ' ', cleanText)
    return cleanText
# Prediction and Category Name
def predict_category(resume_text):
   resume_text= cleanResume(resume_text)
   resume_tfidf = tfidf_vectorizer.transform([resume_text])
    predicted_category = rf_classifier.predict(resume_tfidf)[0]
    return predicted_category
#Example For the same
resume_file = """hr assistant summary hard worker dedicated detailed oriented experienced organized highly motivated many attributes possess ind
predicted_category = predict_category(resume_file)
print("Predicted Category:", predicted_category)
→ Predicted Category: HR
```

#one more example
Example Usage
resume_file = """Objective:

A creative and detail-oriented Designer with a passion for visual communication and brand identity seeking opportunities to leverage design skil

Education:

- Bachelor of Fine Arts in Graphic Design, XYZ College, GPA: 3.7/4.0
- Diploma in Web Design, ABC Institute, GPA: 3.9/4.0

Skills:

- Proficient in Adobe Creative Suite (Photoshop, Illustrator, InDesign)
- Strong understanding of typography, layout, and color theory
- Experience in both print and digital design
- Ability to conceptualize and execute design projects from concept to completion $% \left(1\right) =\left(1\right) \left(1\right)$
- Excellent attention to detail and time management skills

Experience:

Graphic Designer | XYZ Design Studio

- Created visually appealing graphics for various marketing materials, including brochures, flyers, and social media posts
- Collaborated with clients to understand their design needs and deliver creative solutions that align with their brand identity
- Worked closely with the marketing team to ensure consistency in brand messaging across all platforms

Freelance Designer

- Designed logos, branding materials, and website layouts for small businesses and startups
- Managed multiple projects simultaneously while meeting tight deadlines and maintaining quality standards
- Established and maintained strong client relationshins through clear communication and exceptional service