```
In [1]:
```

import pandas as pd
.

import numpy as np

import os

import seaborn as sns

import matplotlib.pyplot as plt

 $from \ sklearn.feature_extraction.text \ import \ TfidfVectorizer$

from sklearn.feature_selection import VarianceThreshold

In [2]:

df = pd.read_csv("/Users/hjk2160@columbia.edu/Desktop/dataset.csv")

In [3]:

df.head()

Out[3]:

																Out	[3]:
	Un na me d: 0	in d e x		Sal ary Esti mat e	Job Description	R at in g	Com pany Nam e	Loc atio n	Head quar ters	Siz e	Fo un de d	Typ e of ow ner shi p	Ind ustr y	Sect or		Com petit ors	as y A p pl
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In [4]:

df.drop('Unnamed: 0', axis=1, inplace=True)

In [5]:

from collections import Counter #항목들 계산해주는 라이브러리 import re #정규표현식임. 쉽게 말해, 내 키워드같은거 어떤 corpus 에서 찾을수 있는지 찾아주는거임.

```
keywords = ['data', 'machine learning', 'analysis', 'statistics',
'research', 'python']
def count keywords (description, keywords):
    description = description.lower()
    keyword counts = {keyword:
len(re.findall(r'\b{}\b'.format(re.escape(keyword)), description)) for
keyword in keywords}
    return keyword counts
                                                                        In [6]:
max keyword occurrence = 0
                                                                        In [7]:
for description in df['Job Description']:
    keyword_counts = count_keywords(description, keywords)
    max keyword occurrence = max(max keyword occurrence,
max(keyword counts.values()))
                                                                        In [8]:
def calculate score(keyword counts, max score):
    total count = sum(keyword counts.values())
    normalized score = 0.1 + 0.9 * (total count / max score) if max score >
0 else 0.1
    return normalized score
```

```
df['Keyword Counts'] = df['Job Description'].apply(lambda desc:
count keywords(desc, keywords))
df['Keyword Score'] = df['Keyword Counts'].apply(lambda counts:
calculate score(counts, max keyword occurrence))
                                                                               In [9]:
df.drop('Job Description', axis=1, inplace=True)
                                                                              In [10]:
df.drop('index',axis=1, inplace=True)
                                                                              In [11]:
# need to convert salary estimate into scores as well
def salary to_score(salary_string):
    salary string = re.sub(r'(\(Glassdoor est.\))|K|\', '',
salary_string).strip()
    numbers = re.findall(r'\d+', salary string)
    if len(numbers) == 2:
         lower bound = int(numbers[0]) * 1000
         upper bound = int(numbers[1]) * 1000
         score = (lower bound + upper bound) / 2
    else:
         score = None
    return score
                                                                              In [12]:
df['Salary Score'] = df['Salary Estimate'].apply(salary to score)
df = df.dropna(subset=['Salary Score'])
max_salary_score = df['Salary_Score'].max()
min salary score = df['Salary Score'].min()
df['Salary_Estimate_Score'] = df['Salary_Score'].apply(lambda x: 0.1 + 0.9
* ((x - min_salary_score) / (max_salary_score - min_salary_score)))
                                                                              In [13]:
df.head()
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```
keywords = ['senior', 'analyst', 'director', 'manager', 'sr', 'lead',
'principal']
def contains keyword(title, keywords):
    title lower = title.lower()
    return any(keyword in title lower for keyword in keywords)
mask = df['Job Title'].apply(contains keyword, keywords=keywords)
df = df[\sim mask]
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df.head(10)
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3	Data Scient ist	\$1 11 K- \$1 81 K (Gl ass do or est.	3. 9	Point 72\n 3.9		Stam ford, CT		20 14	Co mp any - Pri vat e	Inve stm ent Ban king & Ass et Man age men t	anc	Un kn ow n / No n- Ap pli cab le	-1	-1	{'data ': 14, 'mach ine learni ng': 2, 'analy sis'	0.48 0769	146 000. 0	0.6565 22

In [16]:

df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2149 entries, 1 to 3908
Data columns (total 17 columns):

#	Column	Non-Null Count	Dtype
0	Job Title	2149 non-null	object
1	Salary Estimate	2149 non-null	object
2	Rating	2149 non-null	float64
3	Company Name	2149 non-null	object
4	Location	2149 non-null	object
5	Headquarters	2149 non-null	object
6	Size	2149 non-null	object
7	Type of ownership	2149 non-null	object
8	Industry	2149 non-null	object
9	Sector	2149 non-null	object
10	Revenue	2149 non-null	object
11	Competitors	2149 non-null	object
12	Easy Apply	2149 non-null	object
13	Keyword_Counts	2149 non-null	object
14	Keyword_Score	2149 non-null	float64
15	Salary_Score	2149 non-null	float64
16	Salary_Estimate_Score	2149 non-null	float64
d+ vn	es: float64(4) object(13)	

dtypes: float64(4), object(13)

memory usage: 302.2+ KB

df.drop('Size', axis = 1, inplace=True)

df.drop('Easy Apply', axis=1, inplace=True)

df.describe()

	Rating	Keyword_Score	Salary_Score	Salary_Estimate_Score
count	2149.000000	2149.000000	2149.000000	2149.000000
mean	3.265891	0.331161	108511.167985	0.493527
std	1.628358	0.171478	37667.785608	0.163773
min	-1.000000	0.100000	18000.000000	0.100000
25%	3.200000	0.203846	77500.000000	0.358696
50%	3.700000	0.290385	105500.000000	0.480435
75%	4.100000	0.428846	133000.000000	0.600000
max	5.000000	1.138462	225000.000000	1.000000

In [21]:

In [18]:

In [19]:

In [20]:

Out[20]:

df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2149 entries, 1 to 3908
Data columns (total 15 columns):

Column Non-Null Count Dtype

0	Job Title	2149	non-null	object
1	Salary Estimate	2149	non-null	object
2	Rating	2149	non-null	float64
3	Company Name	2149	non-null	object
4	Location	2149	non-null	object
5	Headquarters	2149	non-null	object
6	Type of ownership	2149	non-null	object
7	Industry	2149	non-null	object
8	Sector	2149	non-null	object
9	Revenue	2149	non-null	object
10	Competitors	2149	non-null	object
11	Keyword_Counts	2149	non-null	object
12	Keyword_Score	2149	non-null	float64
13	Salary_Score	2149	non-null	float64
14	Salary_Estimate_Score	2149	non-null	float64

dtypes: float64(4), object(11)

memory usage: 268.6+ KB

EDA 파트 1

In [22]:

#본격적인 EDA

numerical_vars = ['Keyword_Score', 'Salary_Score', 'Salary_Estimate_Score',
'Rating']

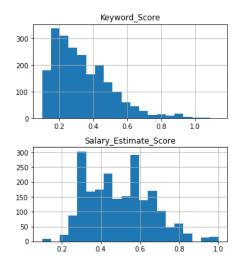
df[numerical_vars].describe().T

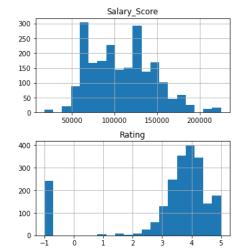
Out[22]:

	cou nt	mean	std	min	25%	50%	75%	max
Keyword_Scor e	214 9.0	0.331161	0.171478	0.1	0.203846	0.290385	0.428846	1.138462
Salary_Score	214 9.0	108511.16 7985			77500.00 0000		133000.00 0000	225000.00 0000
Salary_Estimat e_Score	214 9.0	0.493527	0.163773	0.1	0.358696	0.480435	0.600000	1.000000
Rating	214 9.0	3.265891	1.628358	-1.0	3.200000	3.700000	4.100000	5.000000

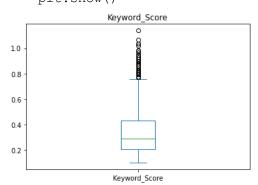
In [23]:

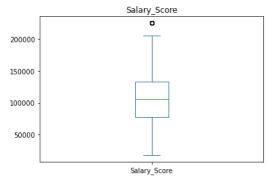
df[numerical_vars].hist(bins=20, figsize=(12,6))
plt.show()

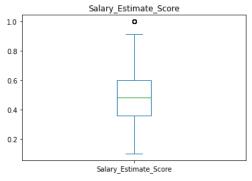




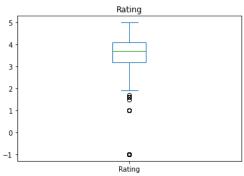
for var in numerical_vars:
 df[var].plot(kind='box')
 plt.title(var)
 plt.show()







In [24]:



Rating		
		In [25]:
df['Job Title'].value_count	s()	
Data Scientist		Out[25]:
Data Engineer		260
Machine Learning Engineer		47
Big Data Engineer		41
Research Scientist		22
Research Sciencisc		22
Scottsdale Data Science Tut	or Johs	1
Mesa Data Science Tutor Job		1
Cloud Data Engineer - Solut		1
Chandler Data Science Tutor		1
	Safety Scientist - UK, Europe or the US	1
Name: Job Title, Length: 11		_
	ar, magratical and a	In [26]:
<pre>value counts = df['Job Titl</pre>	e'].value counts()	m [20].
_	<pre>(value counts[value counts >= 20].index)</pre>	
df = df[mask]		
		In [27]:
<pre>df['Job Title'].value_count</pre>	s()	
		Out[27]:
Data Scientist	274	
Data Engineer	260	
Machine Learning Engineer	47	
Big Data Engineer	41	
Research Scientist	22	
Name: Job Title, dtype: int	64	
1611-1		In [28]:
df['Industry'].value_counts	()	0 . [00]
-1	145	Out[28]:
IT Services	102	
Staffing & Outsourcing	49	
Enterprise Software & Netwo		
Consulting	40	
0011541 01119		
Food & Beverage Manufacturi		
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Casual Restaurants	1	
Grantmaking Foundations	1	
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Name: Industry, Length: 65, dtype: int64
                                                                            In [29]:
df = df[df['Industry'] != '-1']
                                                                            In [30]:
value counts = df['Industry'].value counts()
mask = df['Industry'].isin(value counts[value counts >= 20].index)
df = df[mask]
                                                                            In [31]:
df['Industry'].value counts()
                                                                           Out[31]:
IT Services
                                               102
Staffing & Outsourcing
                                                49
Enterprise Software & Network Solutions
                                                42
Consulting
                                                40
Computer Hardware & Software
                                                38
Internet
                                                32
Advertising & Marketing
                                                24
Name: Industry, dtype: int64
                                                                            In [32]:
def categorize industry(industry):
    if industry == 'Consulting':
        return 'Consulting Industry'
    elif industry in ['IT Services', 'Computer Hardware & Software',
'Internet', 'Enterprise Software & Network Solutions', 'Advertising &
Marketing']:
        return 'Tech Industry'
    else:
        return 'Other Industry'
                                                                            In [33]:
df['categorized industry'] = df['Industry'].apply(categorize industry)
                                                                            In [34]:
df.drop('Competitors', axis=1, inplace=True)
                                                                            In [35]:
df.drop('Sector', axis=1, inplace=True)
                                                                            In [36]:
df.drop('Revenue', axis=1, inplace=True)
                                                                            In [37]:
df.drop('Location', axis=1, inplace=True)
                                                                            In [38]:
df.drop('Headquarters', axis=1, inplace=True)
                                                                            In [39]:
df.head()
                                                                           Out[39]:
       Salar
                         Type
   Iob
                 Compa
           y Rat
                           of Indus Keyword
                                             Keywor Salary
                                                             Salary_Esti
                                                                        categorize
   Titl
                     ny
                                             d_Score | _Score | mate_Score |
        Esti ing
                        owne
                                     _Counts
                                                                        d_industry
                                try
                  Name
     e
       mate
                        rship
1 Data $111
                 Quartet
                              Enter
                                    {'data': 8,
                                             0.41153
                                                     14600
                        Comp
                                                                       Tech
            3.9
                                                            0.656522
O Scie K-
                Health\
                                    'machine
                              prise
                                                     0.0
                                                                       Industry
                        any -
```

ntist \$181

n3.9

Softw

learning':

	Job Titl e	Salar y Esti mate	Rat ing	Compa ny Name	Type of owne rship	Indus try	Keyword _Counts	Keywor d_Score	Salary _Score	Salary_Esti mate_Score	categorize d_industry
		K (Glas sdoor est.)			Privat e	are & Netw ork Soluti ons	4, 'analysis': 				
1 5	Scie	\$111 K- \$181 K (Glas sdoor est.)	3.0	Affinity Solutio ns\n3.0	Comp any - Privat e	Adver tising & Mark eting	{'data': 8, 'machine learning': 2, 'analysis': 	0.34230 8	14600 0.0	0.656522	Tech Industry
2 4	Scie	\$111 K- \$181 K (Glas sdoor est.)	4.4	WITHI N\n4.4	Comp any - Privat e	Adver tising & Mark eting	{'data': 6, 'machine learning': 1, 'analysis': 	0.23846	14600 0.0	0.656522	Tech Industry
6	Scie	\$111 K- \$181 K (Glas sdoor est.)	4.1	Datado g\n4.1	Comp any - Public	ware	{'data': 8, 'machine learning': 2, 'analysis': 	0.34230 8	14600 0.0	0.656522	Tech Industry
3 0	Scie	\$120 K- \$140 K (Glas sdoor est.)	4.3	Caserta \n4.3	Comp any - Privat e	IT Servic es	{'data': 10, 'machine learning': 1, 'analysis'.	0.35961 5	13000 0.0	0.586957	Tech Industry

In [40]:

df.drop("Type of ownership", axis=1, inplace=True)

In [41]:

df.head()

Out[41]:

	Job Title		Rat ing	Compa ny Name	Indus try	Keyword_ Counts	-		Salary_Estim ate_Score	categorized _industry
0	Data	\$181	3.9	Quartet Health\ n3.9	rise Softwa	{'data': 8, 'machine learning': 4, 'analysis':.	0.411538	146000 .0	111 656577	Tech Industry

	Job Title	Salar y Estim ate	Rat	Compa ny Name	Indus try	Keyword_ Counts	Keyword _Score	Salary_ Score	Salary_Estim ate_Score	categorized _industry
		door est.)			Soluti ons					
1 5	Scie	\$111 K- \$181 K (Glass door est.)	3.0		Advert ising & Marke ting	{'data': 8, 'machine learning': 2, 'analysis':.	0.342308	146000 .0	0.656522	Tech Industry
24	Scie	\$111 K- \$181 K (Glass door est.)	4.4	WITHIN \n4.4	Advert ising & Marke ting	{'data': 6, 'machine learning': 1, 'analysis':.	0.238462	146000 .0	0.656522	Tech Industry
6	Scie	\$111 K- \$181 K (Glass door est.)	4.1	Datadog \n4.1	are &	{'data': 8, 'machine learning': 2, 'analysis':.	0.342308	146000 .0	0.656522	Tech Industry
3 0	Scie	\$120 K- \$140 K (Glass door est.)	4.3	Caserta\ n4.3	IT Servic es	{'data': 10, 'machine learning': 1, 'analysis'	0.359615	130000 .0	0.586957	Tech Industry
<pre>In [43]: df.drop('Salary Estimate', axis=1, inplace=True) df.drop('Industry', axis=1, inplace=True)</pre>										
	<pre>custom_colors = ['#339FFF']</pre> In [44]:									

```
In [43]

df.drop('Salary Estimate', axis=1, inplace=True)

df.drop('Industry', axis=1, inplace=True)

custom_colors = ['#339FFF']

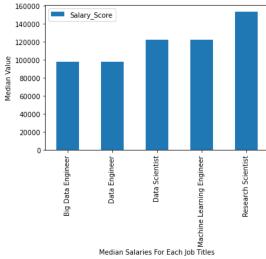
grouped = df.groupby('categorized_industry').agg({'Salary_Score': 'median'}))

fig, ax = plt.subplots()

grouped.plot(kind='bar', ax=ax, color=custom_colors, legend=False)
```

```
ax.set_xlabel('Industry_Salaries', fontsize=12, fontweight='bold')
ax.set_ylabel('Median Value', fontsize=12, fontweight='bold')
ax.tick params(axis='both', which='both', labelsize=10, width=2, length=6)
for spine in ax.spines.values():
     spine.set linewidth(2)
ax.set_xticklabels(ax.get_xticklabels(), rotation=0, fontsize=10)
plt.show()
  120000
  100000
Median Value
   80000
   60000
   40000
   20000
                  Other Industry
Industry_Salaries
        Consulting Industry
                                Tech Industry
                                                                                    In [45]:
```

```
grouped = df.groupby('Job Title').agg({'Salary_Score': 'median'})
fig, ax = plt.subplots()
grouped.plot(kind='bar', ax=ax)
ax.set_xlabel('Median Salaries For Each Job Titles')
ax.set_ylabel('Median Value')
plt.show()
```



In [46]:

df.corr()

Out[46]:

	Rating	Keyword_Score	Salary_Score	Salary_Estimate_Score
Rating	1.000000	0.077334	0.041348	0.041348
Keyword_Score	0.077334	1.000000	-0.016377	-0.016377
Salary_Score	0.041348	-0.016377	1.000000	1.000000
Salary_Estimate_Score	0.041348	-0.016377	1.000000	1.000000

Out[47]:

	Job Title	Rati ng	Compa ny Name	Keyword_Co unts	Keyword_S core	Salary_Sc ore	Salary_Estimate _Score	categorized_in dustry
1 0	Data Scient ist	3.9	Quarte t Health	{'data': 8, 'machine learning': 4, 'analysis':	0.411538	146000.0	0.656522	Tech Industry
1 5	Data Scient ist	3.0	Affinit y Solutio ns	{'data': 8, 'machine learning': 2, 'analysis':	0.342308	146000.0	0.656522	Tech Industry
2 4	Data Scient ist	4.4	WITHI N	{'data': 6, 'machine learning': 1, 'analysis':	0.238462	146000.0	0.656522	Tech Industry
2 6	Data Scient ist	4.1	Datado g	{'data': 8, 'machine learning': 2, 'analysis':	0.342308	146000.0	0.656522	Tech Industry
3 0	Data Scient ist	4.3	Casert a	{'data': 10, 'machine learning': 1, 'analysis'	0.359615	130000.0	0.586957	Tech Industry

In [54]:

df.columns

Out[54]:

두번째 EDA 시작

In [48]:

df.describe()

Out[48]:

	Rating	Keyword_Score	Salary_Score	Salary_Estimate_Score
count	327.000000	327.000000	327.000000	327.000000
mean	3.727217	0.364167	110550.458716	0.502393
std	1.096855	0.153006	33821.432809	0.147050
min	-1.000000	0.117308	43500.000000	0.210870
25%	3.400000	0.255769	83750.000000	0.385870
50%	3.900000	0.342308	109000.000000	0.495652
75%	4.300000	0.463462	132750.000000	0.598913

Ratin	g Keyword_Score	Salary_Score	Salary_Estimate_	Score			
max 5.000000	0.982692	225000.000000	1.000000				
<pre>In [70]: plt.figure(figsize=(10,6)) sns.histplot(data = df, x = "Salary_Score", kde = True, hue = "categorized_industry") plt.xlabel('Salary_Score') plt.ylabel('categorized_industry') Out[70]: Text(0, 0.5, 'categorized industry')</pre>							
40 35 30 30 25 15 10 50000 750	00 100000 125000 Salary 5	150000 175000 2	prized_industry ech Industry consulting Industry ther Industry 2000000 225000				
nl+ figuro/fig	airo=(10 6))			In [65]:			
<pre>plt.figure(figsize=(10,6)) sns.histplot(data = df, x = "Rating", kde = True, hue = "categorized_industry") plt.xlabel('Rating') plt.ylabel('categorized_industry')</pre>							
Text(0, 0.5, '	categorized_in	dustry')		Out[65]:			
categorized_ind Tech Indust Consulting Other Indus 30 20 20 20 20 20 20 20 20 20	ry ndustry						

df.to_csv('updated_DS_jobs.csv', index=False)

10

In [55]:

THEY ARE ALL ENTRY LEVEL DS JOBS!

2 Rating 3

```
In [66]:
```

```
plt.figure(figsize=(10,6))
sns.scatterplot(data=df, x='Rating', y='Salary_Score',
hue='categorized_industry')
plt.title('Rating and Salary')
plt.legend(title='industry', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.show()
```



In []: