

Data Science Salary Prediction

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1 Predicting Data Science Salaries Worldwide

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```
[1]: import random
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
import statsmodels.formula.api as smf
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder
```

1.1 Import dataset CSV file as a dataframe.

```
[2]: df = pd.read_csv(r"Data Science Jobs Salaries.csv")
```

```
[3]: df
```

```
[3]:
```

	work_year	experience_level	employment_type	job_title	\
0	2021e	EN	FT	Data Science Consultant	
1	2020	SE	FT	Data Scientist	
2	2021e	EX	FT	Head of Data Science	
3	2021e	EX	FT	Head of Data	
4	2021e	EN	FT	Machine Learning Engineer	
..	
240	2020	SE	FT	Data Scientist	
241	2021e	MI	FT	Principal Data Scientist	
242	2020	EN	FT	Data Scientist	
243	2020	EN	CT	Business Data Analyst	
244	2021e	SE	FT	Data Science Manager	

	salary	salary_currency	salary_in_usd	employee_residence	remote_ratio	\
0	54000	EUR	64369	DE	50	
1	60000	EUR	68428	GR	100	
2	85000	USD	85000	RU	0	
3	230000	USD	230000	RU	50	
4	125000	USD	125000	US	100	

```

..      ...      ...      ...      ...      ...
240    412000      USD      412000      US      100
241    151000      USD      151000      US      100
242    105000      USD      105000      US      100
243    100000      USD      100000      US      100
244    7000000      INR      94917      IN      50

```

```

      company_location company_size
0                DE            L
1                US            L
2                RU            M
3                RU            L
4                US            S
..      ...      ...
240        US            L
241        US            L
242        US            S
243        US            L
244        IN            L

```

```
[245 rows x 11 columns]
```

2 Data Preprocessing

2.1 Drop unnecessary columns

```
[4]: df.drop(columns=['salary_currency', 'salary'], inplace=True)
```

```
[5]: df = df.dropna()
```

2.2 One-hot encode categorical variables

```
[6]: df = pd.get_dummies(df)
```

```
[7]: for col in df.columns:
      df.rename(columns={col : '_' + col.split('_')[1]}, inplace=True)
df.columns
```

```
[7]: Index(['salary_in_usd', 'remote_ratio', 'work_year_2020', 'work_year_2021e',
          'experience_level_EN', 'experience_level_EX', 'experience_level_MI',
          'experience_level_SE', 'employment_type_CT', 'employment_type_FL',
          ...,
          'company_location_RU', 'company_location_SG', 'company_location_SI',
          'company_location_TR', 'company_location_UA', 'company_location_US',
          'company_location_VN', 'company_size_L', 'company_size_M',
          'company_size_S'],
          dtype=object)
```

```
dtype='object', length=144)
```

2.3 Create n-1 columns for categorical variables to prevent the dummy variable trap. Drop 1 of each.

```
[8]: df.drop(columns=['work_year_2020', 'experience_level_EN', 'employment_type_CT', 'job_title_3D_Computer_Vision_Researcher', 'employee_residence_AE', 'company_location_AE', 'company_size_L'], inplace=True)
```

2.4 Process DataFrame into more consistent, clear categorical variables.

```
[9]: df
```

```
[9]:
```

	salary_in_usd	remote_ratio	work_year_2021e	experience_level_EX	\
0	64369	50	1	0	
1	68428	100	0	0	
2	85000	0	1	1	
3	230000	50	1	1	
4	125000	100	1	0	
..	
240	412000	100	0	0	
241	151000	100	1	0	
242	105000	100	0	0	
243	100000	100	0	0	
244	94917	50	1	0	

	experience_level_MI	experience_level_SE	employment_type_FL	\
0	0	0	0	
1	0	1	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	
..	
240	0	1	0	
241	1	0	0	
242	0	0	0	
243	0	0	0	
244	0	1	0	

	employment_type_FT	employment_type_PT	job_title_AI_Scientist	...	\
0	1	0	0	...	
1	1	0	0	...	
2	1	0	0	...	
3	1	0	0	...	
4	1	0	0	...	
..	

240	1	0	0	...
241	1	0	0	...
242	1	0	0	...
243	0	0	0	...
244	1	0	0	...

	company_location_PT	company_location_RU	company_location_SG	\
0	0	0	0	
1	0	0	0	
2	0	1	0	
3	0	1	0	
4	0	0	0	
..	
240	0	0	0	
241	0	0	0	
242	0	0	0	
243	0	0	0	
244	0	0	0	

	company_location_SI	company_location_TR	company_location-UA	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	
..	
240	0	0	0	
241	0	0	0	
242	0	0	0	
243	0	0	0	
244	0	0	0	

	company_location_US	company_location_VN	company_size_M	company_size_S
0	0	0	0	0
1	1	0	0	0
2	0	0	1	0
3	0	0	0	0
4	1	0	0	1
..
240	1	0	0	0
241	1	0	0	0
242	1	0	0	1
243	1	0	0	0
244	0	0	0	0

[245 rows x 137 columns]

```
[10]: # Unique values of each column
print('COLUMN NULL COUNT')
column_types = {}
for col in df.columns:
    # print(i)
    # print(df[i].unique())
    column_types[col] = type(df[col][0])
    null_count = sum(df[col].isna())
    print(col + ': ', null_count)

    # print('-----')

print('\nCOLUMN DATA TYPE')
for col, type in column_types.items():
    print(col + ': ', type)
```

```
COLUMN NULL COUNT
salary_in_usd: 0
remote_ratio: 0
work_year_2021e: 0
experience_level_EX: 0
experience_level_MI: 0
experience_level_SE: 0
employment_type_FL: 0
employment_type_FT: 0
employment_type_PT: 0
job_title_AI_Scientist: 0
job_title_Applied_Data_Scientist: 0
job_title_Applied_Machine_Learning_Scientist: 0
job_title_BI_Data_Analyst: 0
job_title_Big_Data_Architect: 0
job_title_Big_Data_Engineer: 0
job_title_Business_Data_Analyst: 0
job_title_Cloud_Data_Engineer: 0
job_title_Computer_Vision_Engineer: 0
job_title_Computer_Vision_Software_Engineer: 0
job_title_Data_Analyst: 0
job_title_Data_Analytics_Engineer: 0
job_title_Data_Analytics_Manager: 0
job_title_Data_Architect: 0
job_title_Data_Engineer: 0
job_title_Data_Engineering_Manager: 0
job_title_Data_Science_Consultant: 0
job_title_Data_Science_Engineer: 0
job_title_Data_Science_Manager: 0
job_title_Data_Scientist: 0
job_title_Data_Specialist: 0
```

job_title_Director_of_Data_Engineering: 0
job_title_Director_of_Data_Science: 0
job_title_Finance_Data_Analyst: 0
job_title_Financial_Data_Analyst: 0
job_title_Head_of_Data: 0
job_title_Head_of_Data_Science: 0
job_title_Lead_Data_Analyst: 0
job_title_Lead_Data_Engineer: 0
job_title_Lead_Data_Scientist: 0
job_title_ML_Engineer: 0
job_title_Machine_Learning_Engineer: 0
job_title_Machine_Learning_Infrastructure_Engineer: 0
job_title_Machine_Learning_Scientist: 0
job_title_Manager_Data_Science: 0
job_title_Marketing_Data_Analyst: 0
job_title_Principal_Data_Analyst: 0
job_title_Principal_Data_Engineer: 0
job_title_Principal_Data_Scientist: 0
job_title_Product_Data_Analyst: 0
job_title_Research_Scientist: 0
job_title_Staff_Data_Scientist: 0
employee_residence_AT: 0
employee_residence_BE: 0
employee_residence_BG: 0
employee_residence_BR: 0
employee_residence_CA: 0
employee_residence_CL: 0
employee_residence_CN: 0
employee_residence_CO: 0
employee_residence_DE: 0
employee_residence_DK: 0
employee_residence_ES: 0
employee_residence_FR: 0
employee_residence_GB: 0
employee_residence_GR: 0
employee_residence_HK: 0
employee_residence_HR: 0
employee_residence_HU: 0
employee_residence_IN: 0
employee_residence_IR: 0
employee_residence_IT: 0
employee_residence_JE: 0
employee_residence_JP: 0
employee_residence_KE: 0
employee_residence_LU: 0
employee_residence_MD: 0
employee_residence_MT: 0
employee_residence_MX: 0

employee_residence_NG: 0
employee_residence_NL: 0
employee_residence_NZ: 0
employee_residence_PH: 0
employee_residence_PK: 0
employee_residence_PL: 0
employee_residence_PR: 0
employee_residence_PT: 0
employee_residence_RO: 0
employee_residence_RS: 0
employee_residence_RU: 0
employee_residence_SG: 0
employee_residence_SI: 0
employee_residence_TR: 0
employee_residence_UA: 0
employee_residence_US: 0
employee_residence_VN: 0
company_location_AS: 0
company_location_AT: 0
company_location_BE: 0
company_location_BR: 0
company_location_CA: 0
company_location_CH: 0
company_location_CL: 0
company_location_CN: 0
company_location_CO: 0
company_location_DE: 0
company_location_DK: 0
company_location_ES: 0
company_location_FR: 0
company_location_GB: 0
company_location_GR: 0
company_location_HR: 0
company_location_HU: 0
company_location_IL: 0
company_location_IN: 0
company_location_IR: 0
company_location_IT: 0
company_location_JP: 0
company_location_KE: 0
company_location_LU: 0
company_location_MD: 0
company_location_MT: 0
company_location_MX: 0
company_location_NG: 0
company_location_NL: 0
company_location_NZ: 0
company_location_PK: 0

company_location_PL: 0
company_location_PT: 0
company_location_RU: 0
company_location_SG: 0
company_location_SI: 0
company_location_TR: 0
company_location_UA: 0
company_location_US: 0
company_location_VN: 0
company_size_M: 0
company_size_S: 0

COLUMN DATA TYPE

salary_in_usd: <class 'numpy.int64'>
remote_ratio: <class 'numpy.int64'>
work_year_2021e: <class 'numpy.uint8'>
experience_level_EX: <class 'numpy.uint8'>
experience_level_MI: <class 'numpy.uint8'>
experience_level_SE: <class 'numpy.uint8'>
employment_type_FL: <class 'numpy.uint8'>
employment_type_FT: <class 'numpy.uint8'>
employment_type_PT: <class 'numpy.uint8'>
job_title_AI_Scientist: <class 'numpy.uint8'>
job_title_Applied_Data_Scientist: <class 'numpy.uint8'>
job_title_Applied_Machine_Learning_Scientist: <class 'numpy.uint8'>
job_title_BI_Data_Analyst: <class 'numpy.uint8'>
job_title_Big_Data_Architect: <class 'numpy.uint8'>
job_title_Big_Data_Engineer: <class 'numpy.uint8'>
job_title_Business_Data_Analyst: <class 'numpy.uint8'>
job_title_Cloud_Data_Engineer: <class 'numpy.uint8'>
job_title_Computer_Vision_Engineer: <class 'numpy.uint8'>
job_title_Computer_Vision_Software_Engineer: <class 'numpy.uint8'>
job_title_Data_Analyst: <class 'numpy.uint8'>
job_title_Data_Analytics_Engineer: <class 'numpy.uint8'>
job_title_Data_Analytics_Manager: <class 'numpy.uint8'>
job_title_Data_Architect: <class 'numpy.uint8'>
job_title_Data_Engineer: <class 'numpy.uint8'>
job_title_Data_Engineering_Manager: <class 'numpy.uint8'>
job_title_Data_Science_Consultant: <class 'numpy.uint8'>
job_title_Data_Science_Engineer: <class 'numpy.uint8'>
job_title_Data_Science_Manager: <class 'numpy.uint8'>
job_title_Data_Scientist: <class 'numpy.uint8'>
job_title_Data_Specialist: <class 'numpy.uint8'>
job_title_Director_of_Data_Engineering: <class 'numpy.uint8'>
job_title_Director_of_Data_Science: <class 'numpy.uint8'>
job_title_Finance_Data_Analyst: <class 'numpy.uint8'>
job_title_Financial_Data_Analyst: <class 'numpy.uint8'>
job_title_Head_of_Data: <class 'numpy.uint8'>

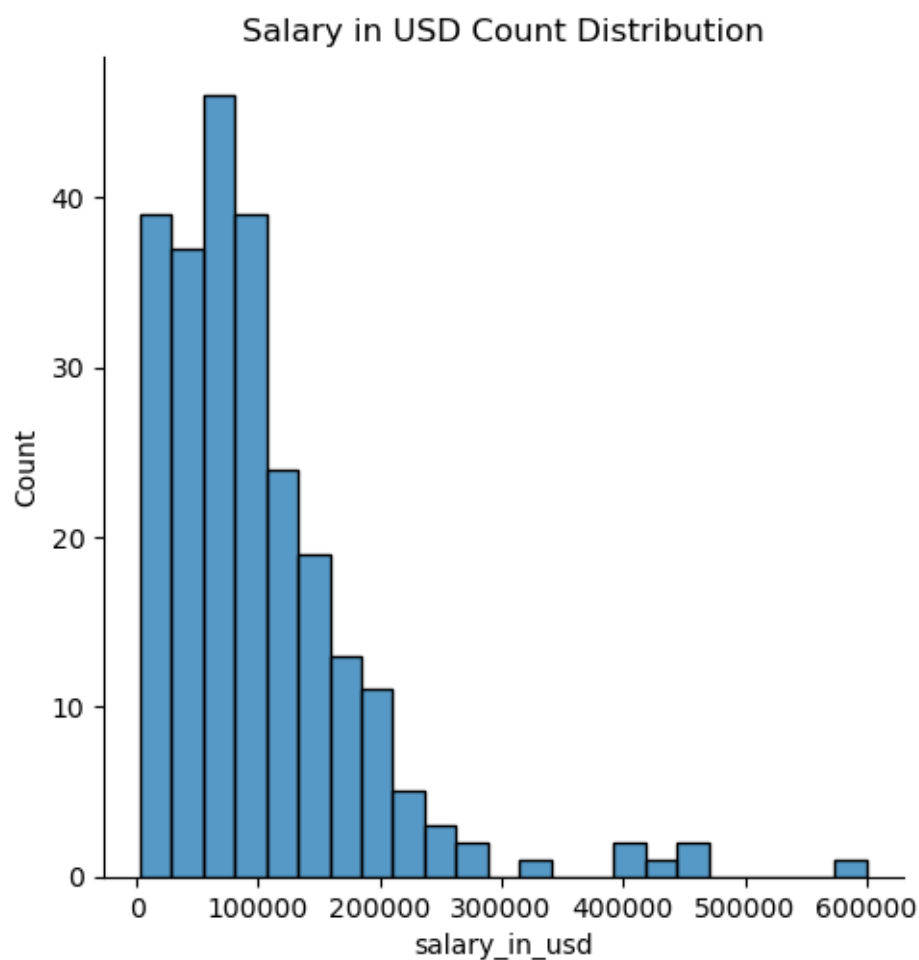
job_title_Head_of_Data_Science: <class 'numpy.uint8'>
job_title_Lead_Data_Analyst: <class 'numpy.uint8'>
job_title_Lead_Data_Engineer: <class 'numpy.uint8'>
job_title_Lead_Data_Scientist: <class 'numpy.uint8'>
job_title_ML_Engineer: <class 'numpy.uint8'>
job_title_Machine_Learning_Engineer: <class 'numpy.uint8'>
job_title_Machine_Learning_Infrastructure_Engineer: <class 'numpy.uint8'>
job_title_Machine_Learning_Scientist: <class 'numpy.uint8'>
job_title_Manager_Data_Science: <class 'numpy.uint8'>
job_title_Marketing_Data_Analyst: <class 'numpy.uint8'>
job_title_Principal_Data_Analyst: <class 'numpy.uint8'>
job_title_Principal_Data_Engineer: <class 'numpy.uint8'>
job_title_Principal_Data_Scientist: <class 'numpy.uint8'>
job_title_Product_Data_Analyst: <class 'numpy.uint8'>
job_title_Research_Scientist: <class 'numpy.uint8'>
job_title_Staff_Data_Scientist: <class 'numpy.uint8'>
employee_residence_AT: <class 'numpy.uint8'>
employee_residence_BE: <class 'numpy.uint8'>
employee_residence_BG: <class 'numpy.uint8'>
employee_residence_BR: <class 'numpy.uint8'>
employee_residence_CA: <class 'numpy.uint8'>
employee_residence_CL: <class 'numpy.uint8'>
employee_residence_CN: <class 'numpy.uint8'>
employee_residence_CO: <class 'numpy.uint8'>
employee_residence_DE: <class 'numpy.uint8'>
employee_residence_DK: <class 'numpy.uint8'>
employee_residence_ES: <class 'numpy.uint8'>
employee_residence_FR: <class 'numpy.uint8'>
employee_residence_GB: <class 'numpy.uint8'>
employee_residence_GR: <class 'numpy.uint8'>
employee_residence_HK: <class 'numpy.uint8'>
employee_residence_HR: <class 'numpy.uint8'>
employee_residence_HU: <class 'numpy.uint8'>
employee_residence_IN: <class 'numpy.uint8'>
employee_residence_IR: <class 'numpy.uint8'>
employee_residence_IT: <class 'numpy.uint8'>
employee_residence_JE: <class 'numpy.uint8'>
employee_residence_JP: <class 'numpy.uint8'>
employee_residence_KE: <class 'numpy.uint8'>
employee_residence_LU: <class 'numpy.uint8'>
employee_residence_MD: <class 'numpy.uint8'>
employee_residence_MT: <class 'numpy.uint8'>
employee_residence_MX: <class 'numpy.uint8'>
employee_residence_NG: <class 'numpy.uint8'>
employee_residence_NL: <class 'numpy.uint8'>
employee_residence_NZ: <class 'numpy.uint8'>
employee_residence_PH: <class 'numpy.uint8'>
employee_residence_PK: <class 'numpy.uint8'>

employee_residence_PL: <class 'numpy.uint8'>
employee_residence_PR: <class 'numpy.uint8'>
employee_residence_PT: <class 'numpy.uint8'>
employee_residence_R0: <class 'numpy.uint8'>
employee_residence_RS: <class 'numpy.uint8'>
employee_residence_RU: <class 'numpy.uint8'>
employee_residence_SG: <class 'numpy.uint8'>
employee_residence_SI: <class 'numpy.uint8'>
employee_residence_TR: <class 'numpy.uint8'>
employee_residence_UA: <class 'numpy.uint8'>
employee_residence_US: <class 'numpy.uint8'>
employee_residence_VN: <class 'numpy.uint8'>
company_location_AS: <class 'numpy.uint8'>
company_location_AT: <class 'numpy.uint8'>
company_location_BE: <class 'numpy.uint8'>
company_location_BR: <class 'numpy.uint8'>
company_location_CA: <class 'numpy.uint8'>
company_location_CH: <class 'numpy.uint8'>
company_location_CL: <class 'numpy.uint8'>
company_location_CN: <class 'numpy.uint8'>
company_location_CO: <class 'numpy.uint8'>
company_location_DE: <class 'numpy.uint8'>
company_location_DK: <class 'numpy.uint8'>
company_location_ES: <class 'numpy.uint8'>
company_location_FR: <class 'numpy.uint8'>
company_location_GB: <class 'numpy.uint8'>
company_location_GR: <class 'numpy.uint8'>
company_location_HR: <class 'numpy.uint8'>
company_location_HU: <class 'numpy.uint8'>
company_location_IL: <class 'numpy.uint8'>
company_location_IN: <class 'numpy.uint8'>
company_location_IR: <class 'numpy.uint8'>
company_location_IT: <class 'numpy.uint8'>
company_location_JP: <class 'numpy.uint8'>
company_location_KE: <class 'numpy.uint8'>
company_location_LU: <class 'numpy.uint8'>
company_location_MD: <class 'numpy.uint8'>
company_location_MT: <class 'numpy.uint8'>
company_location_MX: <class 'numpy.uint8'>
company_location_NG: <class 'numpy.uint8'>
company_location_NL: <class 'numpy.uint8'>
company_location_NZ: <class 'numpy.uint8'>
company_location_PK: <class 'numpy.uint8'>
company_location_PL: <class 'numpy.uint8'>
company_location_PT: <class 'numpy.uint8'>
company_location_RU: <class 'numpy.uint8'>
company_location_SG: <class 'numpy.uint8'>
company_location_SI: <class 'numpy.uint8'>

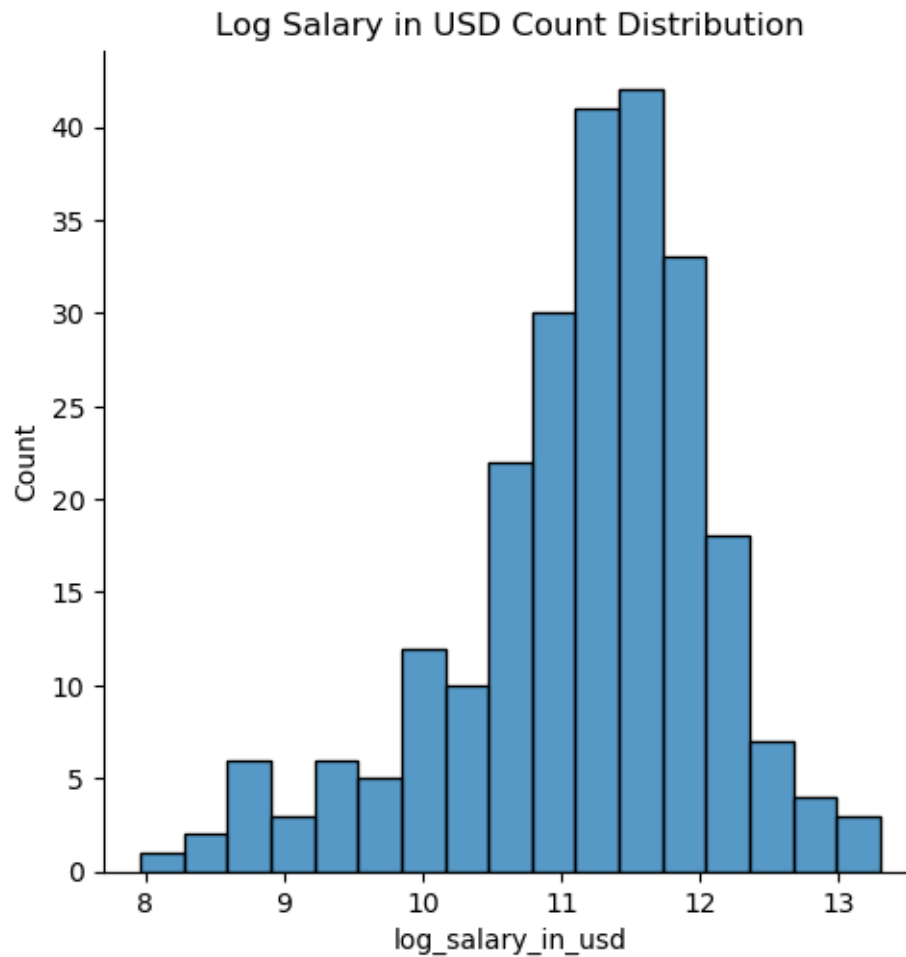
```
company_location_TR: <class 'numpy.uint8'>
company_location_UA: <class 'numpy.uint8'>
company_location_US: <class 'numpy.uint8'>
company_location_VN: <class 'numpy.uint8'>
company_size_M: <class 'numpy.uint8'>
company_size_S: <class 'numpy.uint8'>
```

2.5 Split dataset into train and test sets.

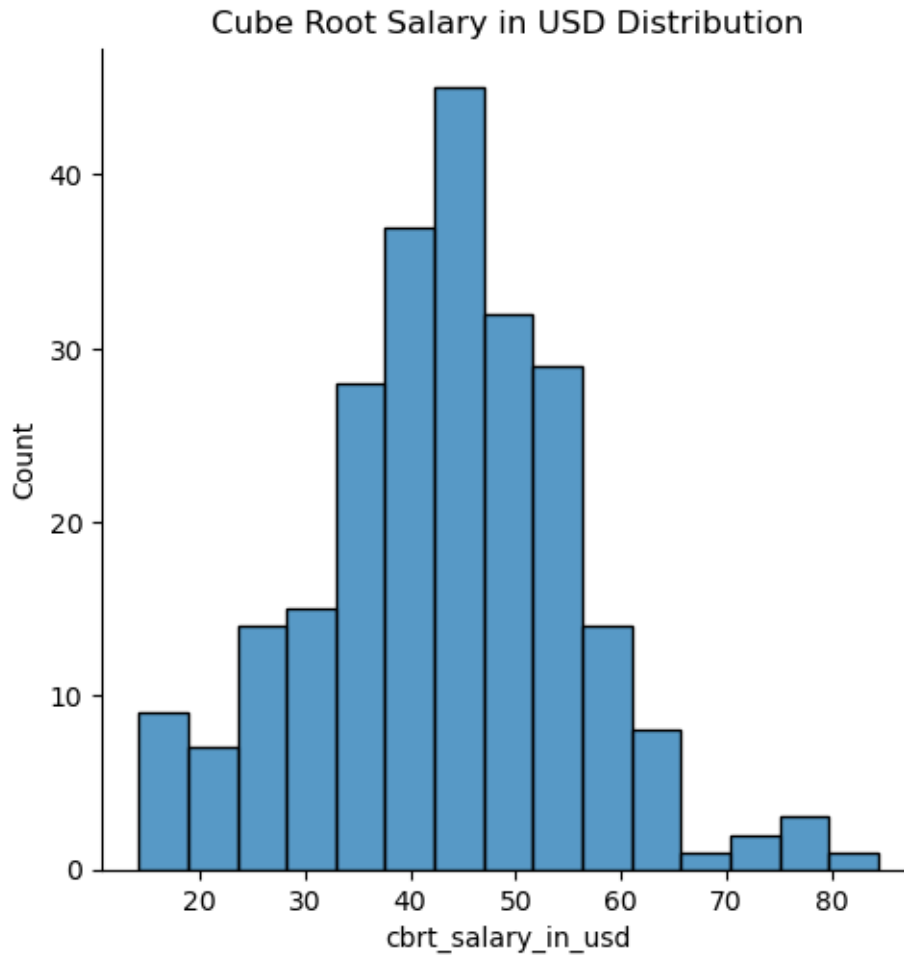
```
[10]: sns.displot(df, x='salary_in_usd')
plt.title('Salary in USD Count Distribution')
plt.show();
```



```
[11]: df['log_salary_in_usd'] = np.log(df['salary_in_usd'].values)
sns.displot(df, x='log_salary_in_usd')
plt.title('Log Salary in USD Count Distribution');
```



```
[12]: df['cbrt_salary_in_usd'] = np.array(df['salary_in_usd'].values)**(1/3) # Cube Root
sns.displot(data=df, x='cbrt_salary_in_usd')
plt.title('Cube Root Salary in USD Distribution')
plt.show();
```



```
[14]: # train test split
df_train, df_test = train_test_split(df, test_size=0.3, random_state=88)
df_train_y_actual = df_train['cbrt_salary_in_usd']
df_test_y_actual = df_test['cbrt_salary_in_usd']
df_train.shape, df_test.shape
```

```
[14]: ((171, 138), (74, 138))
```

```
[15]: # target_column = 'log_salary_in_usd'
target_column = 'cbrt_salary_in_usd'
feature_columns = [col for col in df_train.columns if col != target_column]
all_features = ' + '.join(feature_columns)
all_features
linreg = smf.ols(formula = target_column + ' ~ ' + all_features,
                  data = df_train).fit()
print(linreg.summary())
linreg.summary()
```

OLS Regression Results

```

=====
Dep. Variable:      cbrt_salary_in_usd      R-squared:                0.981
Model:              OLS                    Adj. R-squared:           0.954
Method:             Least Squares          F-statistic:             36.26
Date:               Mon, 08 May 2023       Prob (F-statistic):      1.09e-37
Time:               01:04:21              Log-Likelihood:          -341.37
No. Observations:   171                  AIC:                     884.7
Df Residuals:       70                   BIC:                     1202.
Df Model:           100
Covariance Type:    nonrobust
=====

```

				coef	std err
t	P> t	[0.025	0.975]		

Intercept				21.1044	4.810
4.388	0.000	11.512	30.697		
salary_in_usd				9.947e-05	6.4e-06
15.532	0.000	8.67e-05	0.000		
remote_ratio				-0.0072	0.010
-0.746	0.458	-0.026	0.012		
work_year_2021e				-0.0626	0.707
-0.089	0.930	-1.473	1.348		
experience_level_EX				2.7895	2.897
0.963	0.339	-2.989	8.568		
experience_level_MI				0.8204	0.860
0.954	0.343	-0.894	2.535		
experience_level_SE				2.0696	1.071
1.932	0.057	-0.067	4.206		
employment_type_FL				-7.3516	6.302
-1.167	0.247	-19.920	5.217		
employment_type_FT				6.9583	4.491
1.549	0.126	-1.999	15.916		
employment_type_PT				7.5817	5.592
1.356	0.179	-3.571	18.734		
job_title_AI_Scientist				1.6816	2.945
0.571	0.570	-4.193	7.556		
job_title_Applied_Data_Scientist				7.3678	5.312
1.387	0.170	-3.227	17.963		
job_title_Applied_Machine_Learning_Scientist				-2.9716	3.282
-0.906	0.368	-9.517	3.574		
job_title_BI_Data_Analyst				1.5084	3.069
0.491	0.625	-4.613	7.630		
job_title_Big_Data_Architect				-0.0997	3.425
-0.029	0.977	-6.931	6.732		
job_title_Big_Data_Engineer				-1.1441	2.279

-0.502	0.617	-5.690	3.402		
job_title_Business_Data_Analyst				-2.017e-11	2.29e-11
-0.882	0.381	-6.58e-11	2.54e-11		
job_title_Cloud_Data_Engineer				-1.663e-12	4.04e-12
-0.412	0.682	-9.72e-12	6.39e-12		
job_title_Computer_Vision_Engineer				0.1228	3.424
0.036	0.971	-6.706	6.951		
job_title_Computer_Vision_Software_Engineer				1.0060	2.969
0.339	0.736	-4.916	6.928		
job_title_Data_Analyst				-0.4791	1.190
-0.403	0.688	-2.852	1.894		
job_title_Data_Analytics_Engineer				1.3717	2.883
0.476	0.636	-4.378	7.121		
job_title_Data_Analytics_Manager				2.0740	2.921
0.710	0.480	-3.751	7.900		
job_title_Data_Architect				2.9565	2.843
1.040	0.302	-2.715	8.628		
job_title_Data_Engineer				1.8334	0.868
2.111	0.038	0.102	3.565		
job_title_Data_Engineering_Manager				1.7737	2.353
0.754	0.453	-2.919	6.467		
job_title_Data_Science_Consultant				-0.2006	1.586
-0.126	0.900	-3.363	2.962		
job_title_Data_Science_Engineer				1.1235	2.146
0.524	0.602	-3.156	5.403		
job_title_Data_Science_Manager				5.5431	1.558
3.559	0.001	2.437	8.650		
job_title_Data_Scientist				0.4854	0.844
0.575	0.567	-1.199	2.169		
job_title_Data_Specialist				1.5853	2.816
0.563	0.575	-4.032	7.203		
job_title_Director_of_Data_Engineering				2.3658	2.102
1.125	0.264	-1.827	6.558		
job_title_Director_of_Data_Science				1.5090	2.599
0.581	0.563	-3.674	6.692		
job_title_Finance_Data_Analyst				-2.6147	2.993
-0.874	0.385	-8.585	3.355		
job_title_Financial_Data_Analyst				-3.7327	3.383
-1.103	0.274	-10.479	3.014		
job_title_Head_of_Data				-4.0535	3.443
-1.177	0.243	-10.920	2.813		
job_title_Head_of_Data_Science				1.571e-12	2.61e-12
0.602	0.549	-3.63e-12	6.77e-12		
job_title_Lead_Data_Analyst				-0.5254	2.201
-0.239	0.812	-4.916	3.865		
job_title_Lead_Data_Engineer				0.9303	2.127
0.437	0.663	-3.312	5.173		
job_title_Lead_Data_Scientist				2.4796	2.972

0.834	0.407	-3.449	8.408		
job_title_ML_Engineer				1.9840	2.958
0.671	0.505	-3.916	7.885		
job_title_Machine_Learning_Engineer				1.5101	1.023
1.477	0.144	-0.529	3.550		
job_title_Machine_Learning_Infrastructure_Engineer				3.2782	2.880
1.138	0.259	-2.465	9.021		
job_title_Machine_Learning_Scientist				3.5321	3.386
1.043	0.301	-3.222	10.286		
job_title_Manager_Data_Science				1.2411	2.830
0.439	0.662	-4.404	6.886		
job_title_Marketing_Data_Analyst				2.8309	2.806
1.009	0.317	-2.766	8.428		
job_title_Principal_Data_Analyst				3.1727	2.885
1.100	0.275	-2.580	8.926		
job_title_Principal_Data_Engineer				-12.9105	4.506
-2.865	0.006	-21.898	-3.923		
job_title_Principal_Data_Scientist				3.4608	1.562
2.216	0.030	0.346	6.576		
job_title_Product_Data_Analyst				-8.5013	3.114
-2.730	0.008	-14.712	-2.290		
job_title_Research_Scientist				-0.3904	1.632
-0.239	0.812	-3.644	2.864		
job_title_Staff_Data_Scientist				5.487e-12	6.22e-12
0.882	0.381	-6.91e-12	1.79e-11		
employee_residence_AT				5.4057	5.513
0.981	0.330	-5.590	16.401		
employee_residence_BE				3.8739	1.555
2.491	0.015	0.773	6.975		
employee_residence_BG				4.3732	3.261
1.341	0.184	-2.130	10.877		
employee_residence_BR				-9.7715	5.431
-1.799	0.076	-20.602	1.059		
employee_residence_CA				7.6433	4.696
1.628	0.108	-1.722	17.008		
employee_residence_CL				-1.066e-12	9.75e-13
-1.093	0.278	-3.01e-12	8.79e-13		
employee_residence_CN				-1.6487	2.402
-0.686	0.495	-6.440	3.143		
employee_residence_CO				-0.9162	1.516
-0.604	0.548	-3.940	2.108		
employee_residence_DE				5.2784	3.439
1.535	0.129	-1.580	12.137		
employee_residence_DK				-0.6969	2.652
-0.263	0.794	-5.986	4.593		
employee_residence_ES				4.1136	2.727
1.508	0.136	-1.326	9.553		
employee_residence_FR				4.0950	2.514

1.629	0.108	-0.918	9.108		
employee_residence_GB				-4.1012	2.972
-1.380	0.172	-10.029	1.826		
employee_residence_GR				1.5076	3.174
0.475	0.636	-4.822	7.837		
employee_residence_HK				3.739e-12	3.78e-12
0.990	0.326	-3.79e-12	1.13e-11		
employee_residence_HR				0.5372	1.522
0.353	0.725	-2.498	3.572		
employee_residence_HU				-3.0715	4.195
-0.732	0.467	-11.439	5.296		
employee_residence_IN				-9.3334	2.367
-3.943	0.000	-14.054	-4.613		
employee_residence_IR				1.732e-12	1.67e-12
1.035	0.304	-1.6e-12	5.07e-12		
employee_residence_IT				6.1201	3.425
1.787	0.078	-0.710	12.950		
employee_residence_JE				5.2538	2.466
2.131	0.037	0.336	10.172		
employee_residence_JP				2.5501	1.000
2.550	0.013	0.555	4.545		
employee_residence_KE				-3.9494	2.099
-1.882	0.064	-8.136	0.237		
employee_residence_LU				-1.202e-12	1.31e-12
-0.921	0.360	-3.81e-12	1.4e-12		
employee_residence_MD				-1.2283	1.815
-0.677	0.501	-4.848	2.392		
employee_residence_MT				-1.2776	1.433
-0.892	0.376	-4.135	1.580		
employee_residence_MX				-4.2061	1.150
-3.659	0.000	-6.499	-1.913		
employee_residence_NG				-0.5572	1.041
-0.535	0.594	-2.634	1.520		
employee_residence_NL				7.1599	4.622
1.549	0.126	-2.058	16.378		
employee_residence_NZ				-1.386e-12	1.48e-12
-0.936	0.352	-4.34e-12	1.57e-12		
employee_residence_PH				0.6806	3.022
0.225	0.822	-5.347	6.708		
employee_residence_PK				-9.0381	5.540
-1.632	0.107	-20.087	2.010		
employee_residence_PL				-0.7609	2.618
-0.291	0.772	-5.983	4.461		
employee_residence_PR				5.8451	3.493
1.673	0.099	-1.122	12.813		
employee_residence_PT				2.4410	3.807
0.641	0.524	-5.152	10.034		
employee_residence_RO				-6.2619	3.747

-1.671	0.099	-13.736	1.212		
employee_residence_RS				-1.1814	4.679
-0.253	0.801	-10.513	8.150		
employee_residence_RU				16.0680	5.896
2.725	0.008	4.309	27.827		
employee_residence_SG				5.1570	1.522
3.389	0.001	2.122	8.192		
employee_residence_SI				-1.6038	1.472
-1.090	0.280	-4.539	1.331		
employee_residence_TR				-1.7678	1.120
-1.578	0.119	-4.003	0.467		
employee_residence_UA				-2.6728	1.488
-1.796	0.077	-5.640	0.295		
employee_residence_US				4.8100	1.480
3.251	0.002	1.859	7.761		
employee_residence_VN				-7.7642	4.263
-1.821	0.073	-16.266	0.737		
company_location_AS				5.6315	4.870
1.156	0.251	-4.081	15.344		
company_location_AT				0.7221	4.493
0.161	0.873	-8.239	9.684		
company_location_BE				3.8739	1.555
2.491	0.015	0.773	6.975		
company_location_BR				4.7085	6.168
0.763	0.448	-7.593	17.010		
company_location_CA				0.7494	4.269
0.176	0.861	-7.765	9.264		
company_location_CH				-0.0416	4.015
-0.010	0.992	-8.049	7.966		
company_location_CL				5.331e-18	2.13e-17
0.251	0.803	-3.71e-17	4.77e-17		
company_location_CN				3.6050	1.456
2.476	0.016	0.701	6.509		
company_location_CO				-0.9162	1.516
-0.604	0.548	-3.940	2.108		
company_location_DE				0.3665	3.602
0.102	0.919	-6.818	7.551		
company_location_DK				2.1340	1.939
1.100	0.275	-1.734	6.002		
company_location_ES				-1.3657	3.140
-0.435	0.665	-7.629	4.897		
company_location_FR				-0.8660	2.490
-0.348	0.729	-5.833	4.101		
company_location_GB				10.4463	2.778
3.760	0.000	4.905	15.988		
company_location_GR				1.1235	2.146
0.524	0.602	-3.156	5.403		
company_location_HR				0.5372	1.522

0.353	0.725	-2.498	3.572		
company_location_HU				0	0
nan	nan	0	0		
company_location_IL				5.1570	1.522
3.389	0.001	2.122	8.192		
company_location_IN				7.3119	2.343
3.120	0.003	2.638	11.985		
company_location_IR				0	0
nan	nan	0	0		
company_location_IT				0	0
nan	nan	0	0		
company_location_JP				2.5501	1.000
2.550	0.013	0.555	4.545		
company_location_KE				-3.9494	2.099
-1.882	0.064	-8.136	0.237		
company_location_LU				0	0
nan	nan	0	0		
company_location_MD				-1.2283	1.815
-0.677	0.501	-4.848	2.392		
company_location_MT				-1.2776	1.433
-0.892	0.376	-4.135	1.580		
company_location_MX				-4.2061	1.150
-3.659	0.000	-6.499	-1.913		
company_location_NG				-0.5572	1.041
-0.535	0.594	-2.634	1.520		
company_location_NL				-4.6550	5.471
-0.851	0.398	-15.567	6.257		
company_location_NZ				0	0
nan	nan	0	0		
company_location_PK				0.1987	6.375
0.031	0.975	-12.515	12.913		
company_location_PL				0.5999	2.935
0.204	0.839	-5.255	6.455		
company_location_PT				1.5683	5.002
0.314	0.755	-8.408	11.544		
company_location_RU				-4.0535	3.443
-1.177	0.243	-10.920	2.813		
company_location_SG				0	0
nan	nan	0	0		
company_location_SI				-1.6038	1.472
-1.090	0.280	-4.539	1.331		
company_location_TR				-1.7678	1.120
-1.578	0.119	-4.003	0.467		
company_location_UA				-2.6728	1.488
-1.796	0.077	-5.640	0.295		
company_location_US				2.6906	1.394
1.931	0.058	-0.089	5.470		
company_location_VN				-3.7090	4.790

-0.774	0.441	-13.263	5.845		
company_size_M				-1.5362	0.860
-1.785	0.079	-3.252	0.180		
company_size_S				-0.8028	0.962
-0.835	0.407	-2.721	1.116		

Omnibus:	24.077	Durbin-Watson:	2.067
Prob(Omnibus):	0.000	Jarque-Bera (JB):	67.549
Skew:	-0.519	Prob(JB):	2.15e-15
Kurtosis:	5.899	Cond. No.	5.87e+20

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The smallest eigenvalue is 9.49e-30. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

```
[15]: <class 'statsmodels.iolib.summary.Summary'>
      """
```

```

                                OLS Regression Results
=====
Dep. Variable:      cbrt_salary_in_usd      R-squared:                0.981
Model:                OLS      Adj. R-squared:            0.954
Method:              Least Squares      F-statistic:             36.26
Date:                Mon, 08 May 2023      Prob (F-statistic):       1.09e-37
Time:                01:04:21      Log-Likelihood:          -341.37
No. Observations:      171      AIC:                     884.7
Df Residuals:          70      BIC:                     1202.
Df Model:              100
Covariance Type:      nonrobust
=====
=====
                                coef      std err
t      P>|t|      [0.025      0.975]
-----
Intercept                                21.1044      4.810
4.388      0.000      11.512      30.697
salary_in_usd                        9.947e-05      6.4e-06
15.532      0.000      8.67e-05      0.000
remote_ratio                         -0.0072      0.010
-0.746      0.458      -0.026      0.012
work_year_2021e                      -0.0626      0.707
-0.089      0.930      -1.473      1.348
experience_level_EX                    2.7895      2.897
0.963      0.339      -2.989      8.568

```

experience_level_MI				0.8204	0.860
0.954	0.343	-0.894	2.535		
experience_level_SE				2.0696	1.071
1.932	0.057	-0.067	4.206		
employment_type_FL				-7.3516	6.302
-1.167	0.247	-19.920	5.217		
employment_type_FT				6.9583	4.491
1.549	0.126	-1.999	15.916		
employment_type_PT				7.5817	5.592
1.356	0.179	-3.571	18.734		
job_title_AI_Scientist				1.6816	2.945
0.571	0.570	-4.193	7.556		
job_title_Applied_Data_Scientist				7.3678	5.312
1.387	0.170	-3.227	17.963		
job_title_Applied_Machine_Learning_Scientist				-2.9716	3.282
-0.906	0.368	-9.517	3.574		
job_title_BI_Data_Analyst				1.5084	3.069
0.491	0.625	-4.613	7.630		
job_title_Big_Data_Architect				-0.0997	3.425
-0.029	0.977	-6.931	6.732		
job_title_Big_Data_Engineer				-1.1441	2.279
-0.502	0.617	-5.690	3.402		
job_title_Business_Data_Analyst				-2.017e-11	2.29e-11
-0.882	0.381	-6.58e-11	2.54e-11		
job_title_Cloud_Data_Engineer				-1.663e-12	4.04e-12
-0.412	0.682	-9.72e-12	6.39e-12		
job_title_Computer_Vision_Engineer				0.1228	3.424
0.036	0.971	-6.706	6.951		
job_title_Computer_Vision_Software_Engineer				1.0060	2.969
0.339	0.736	-4.916	6.928		
job_title_Data_Analyst				-0.4791	1.190
-0.403	0.688	-2.852	1.894		
job_title_Data_Analytics_Engineer				1.3717	2.883
0.476	0.636	-4.378	7.121		
job_title_Data_Analytics_Manager				2.0740	2.921
0.710	0.480	-3.751	7.900		
job_title_Data_Architect				2.9565	2.843
1.040	0.302	-2.715	8.628		
job_title_Data_Engineer				1.8334	0.868
2.111	0.038	0.102	3.565		
job_title_Data_Engineering_Manager				1.7737	2.353
0.754	0.453	-2.919	6.467		
job_title_Data_Science_Consultant				-0.2006	1.586
-0.126	0.900	-3.363	2.962		
job_title_Data_Science_Engineer				1.1235	2.146
0.524	0.602	-3.156	5.403		
job_title_Data_Science_Manager				5.5431	1.558

3.559	0.001	2.437	8.650		
job_title_Data_Scientist				0.4854	0.844
0.575	0.567	-1.199	2.169		
job_title_Data_Specialist				1.5853	2.816
0.563	0.575	-4.032	7.203		
job_title_Director_of_Data_Engineering				2.3658	2.102
1.125	0.264	-1.827	6.558		
job_title_Director_of_Data_Science				1.5090	2.599
0.581	0.563	-3.674	6.692		
job_title_Finance_Data_Analyst				-2.6147	2.993
-0.874	0.385	-8.585	3.355		
job_title_Financial_Data_Analyst				-3.7327	3.383
-1.103	0.274	-10.479	3.014		
job_title_Head_of_Data				-4.0535	3.443
-1.177	0.243	-10.920	2.813		
job_title_Head_of_Data_Science				1.571e-12	2.61e-12
0.602	0.549	-3.63e-12	6.77e-12		
job_title_Lead_Data_Analyst				-0.5254	2.201
-0.239	0.812	-4.916	3.865		
job_title_Lead_Data_Engineer				0.9303	2.127
0.437	0.663	-3.312	5.173		
job_title_Lead_Data_Scientist				2.4796	2.972
0.834	0.407	-3.449	8.408		
job_title_ML_Engineer				1.9840	2.958
0.671	0.505	-3.916	7.885		
job_title_Machine_Learning_Engineer				1.5101	1.023
1.477	0.144	-0.529	3.550		
job_title_Machine_Learning_Infrastructure_Engineer				3.2782	2.880
1.138	0.259	-2.465	9.021		
job_title_Machine_Learning_Scientist				3.5321	3.386
1.043	0.301	-3.222	10.286		
job_title_Manager_Data_Science				1.2411	2.830
0.439	0.662	-4.404	6.886		
job_title_Marketing_Data_Analyst				2.8309	2.806
1.009	0.317	-2.766	8.428		
job_title_Principal_Data_Analyst				3.1727	2.885
1.100	0.275	-2.580	8.926		
job_title_Principal_Data_Engineer				-12.9105	4.506
-2.865	0.006	-21.898	-3.923		
job_title_Principal_Data_Scientist				3.4608	1.562
2.216	0.030	0.346	6.576		
job_title_Product_Data_Analyst				-8.5013	3.114
-2.730	0.008	-14.712	-2.290		
job_title_Research_Scientist				-0.3904	1.632
-0.239	0.812	-3.644	2.864		
job_title_Staff_Data_Scientist				5.487e-12	6.22e-12
0.882	0.381	-6.91e-12	1.79e-11		

employee_residence_AT				5.4057	5.513
0.981	0.330	-5.590	16.401		
employee_residence_BE				3.8739	1.555
2.491	0.015	0.773	6.975		
employee_residence_BG				4.3732	3.261
1.341	0.184	-2.130	10.877		
employee_residence_BR				-9.7715	5.431
-1.799	0.076	-20.602	1.059		
employee_residence_CA				7.6433	4.696
1.628	0.108	-1.722	17.008		
employee_residence_CL				-1.066e-12	9.75e-13
-1.093	0.278	-3.01e-12	8.79e-13		
employee_residence_CN				-1.6487	2.402
-0.686	0.495	-6.440	3.143		
employee_residence_CO				-0.9162	1.516
-0.604	0.548	-3.940	2.108		
employee_residence_DE				5.2784	3.439
1.535	0.129	-1.580	12.137		
employee_residence_DK				-0.6969	2.652
-0.263	0.794	-5.986	4.593		
employee_residence_ES				4.1136	2.727
1.508	0.136	-1.326	9.553		
employee_residence_FR				4.0950	2.514
1.629	0.108	-0.918	9.108		
employee_residence_GB				-4.1012	2.972
-1.380	0.172	-10.029	1.826		
employee_residence_GR				1.5076	3.174
0.475	0.636	-4.822	7.837		
employee_residence_HK				3.739e-12	3.78e-12
0.990	0.326	-3.79e-12	1.13e-11		
employee_residence_HR				0.5372	1.522
0.353	0.725	-2.498	3.572		
employee_residence_HU				-3.0715	4.195
-0.732	0.467	-11.439	5.296		
employee_residence_IN				-9.3334	2.367
-3.943	0.000	-14.054	-4.613		
employee_residence_IR				1.732e-12	1.67e-12
1.035	0.304	-1.6e-12	5.07e-12		
employee_residence_IT				6.1201	3.425
1.787	0.078	-0.710	12.950		
employee_residence_JE				5.2538	2.466
2.131	0.037	0.336	10.172		
employee_residence_JP				2.5501	1.000
2.550	0.013	0.555	4.545		
employee_residence_KE				-3.9494	2.099
-1.882	0.064	-8.136	0.237		
employee_residence_LU				-1.202e-12	1.31e-12

-0.921	0.360	-3.81e-12	1.4e-12		
employee_residence_MD				-1.2283	1.815
-0.677	0.501	-4.848	2.392		
employee_residence_MT				-1.2776	1.433
-0.892	0.376	-4.135	1.580		
employee_residence_MX				-4.2061	1.150
-3.659	0.000	-6.499	-1.913		
employee_residence_NG				-0.5572	1.041
-0.535	0.594	-2.634	1.520		
employee_residence_NL				7.1599	4.622
1.549	0.126	-2.058	16.378		
employee_residence_NZ				-1.386e-12	1.48e-12
-0.936	0.352	-4.34e-12	1.57e-12		
employee_residence_PH				0.6806	3.022
0.225	0.822	-5.347	6.708		
employee_residence_PK				-9.0381	5.540
-1.632	0.107	-20.087	2.010		
employee_residence_PL				-0.7609	2.618
-0.291	0.772	-5.983	4.461		
employee_residence_PR				5.8451	3.493
1.673	0.099	-1.122	12.813		
employee_residence_PT				2.4410	3.807
0.641	0.524	-5.152	10.034		
employee_residence_RO				-6.2619	3.747
-1.671	0.099	-13.736	1.212		
employee_residence_RS				-1.1814	4.679
-0.253	0.801	-10.513	8.150		
employee_residence_RU				16.0680	5.896
2.725	0.008	4.309	27.827		
employee_residence_SG				5.1570	1.522
3.389	0.001	2.122	8.192		
employee_residence_SI				-1.6038	1.472
-1.090	0.280	-4.539	1.331		
employee_residence_TR				-1.7678	1.120
-1.578	0.119	-4.003	0.467		
employee_residence_UA				-2.6728	1.488
-1.796	0.077	-5.640	0.295		
employee_residence_US				4.8100	1.480
3.251	0.002	1.859	7.761		
employee_residence_VN				-7.7642	4.263
-1.821	0.073	-16.266	0.737		
company_location_AS				5.6315	4.870
1.156	0.251	-4.081	15.344		
company_location_AT				0.7221	4.493
0.161	0.873	-8.239	9.684		
company_location_BE				3.8739	1.555
2.491	0.015	0.773	6.975		

company_location_BR				4.7085	6.168
0.763	0.448	-7.593	17.010		
company_location_CA				0.7494	4.269
0.176	0.861	-7.765	9.264		
company_location_CH				-0.0416	4.015
-0.010	0.992	-8.049	7.966		
company_location_CL				5.331e-18	2.13e-17
0.251	0.803	-3.71e-17	4.77e-17		
company_location_CN				3.6050	1.456
2.476	0.016	0.701	6.509		
company_location_CO				-0.9162	1.516
-0.604	0.548	-3.940	2.108		
company_location_DE				0.3665	3.602
0.102	0.919	-6.818	7.551		
company_location_DK				2.1340	1.939
1.100	0.275	-1.734	6.002		
company_location_ES				-1.3657	3.140
-0.435	0.665	-7.629	4.897		
company_location_FR				-0.8660	2.490
-0.348	0.729	-5.833	4.101		
company_location_GB				10.4463	2.778
3.760	0.000	4.905	15.988		
company_location_GR				1.1235	2.146
0.524	0.602	-3.156	5.403		
company_location_HR				0.5372	1.522
0.353	0.725	-2.498	3.572		
company_location_HU				0	0
nan	nan	0	0		
company_location_IL				5.1570	1.522
3.389	0.001	2.122	8.192		
company_location_IN				7.3119	2.343
3.120	0.003	2.638	11.985		
company_location_IR				0	0
nan	nan	0	0		
company_location_IT				0	0
nan	nan	0	0		
company_location_JP				2.5501	1.000
2.550	0.013	0.555	4.545		
company_location_KE				-3.9494	2.099
-1.882	0.064	-8.136	0.237		
company_location_LU				0	0
nan	nan	0	0		
company_location_MD				-1.2283	1.815
-0.677	0.501	-4.848	2.392		
company_location_MT				-1.2776	1.433
-0.892	0.376	-4.135	1.580		
company_location_MX				-4.2061	1.150

-3.659	0.000	-6.499	-1.913		
company_location_NG				-0.5572	1.041
-0.535	0.594	-2.634	1.520		
company_location_NL				-4.6550	5.471
-0.851	0.398	-15.567	6.257		
company_location_NZ				0	0
nan	nan	0	0		
company_location_PK				0.1987	6.375
0.031	0.975	-12.515	12.913		
company_location_PL				0.5999	2.935
0.204	0.839	-5.255	6.455		
company_location_PT				1.5683	5.002
0.314	0.755	-8.408	11.544		
company_location_RU				-4.0535	3.443
-1.177	0.243	-10.920	2.813		
company_location_SG				0	0
nan	nan	0	0		
company_location_SI				-1.6038	1.472
-1.090	0.280	-4.539	1.331		
company_location_TR				-1.7678	1.120
-1.578	0.119	-4.003	0.467		
company_location_UA				-2.6728	1.488
-1.796	0.077	-5.640	0.295		
company_location_US				2.6906	1.394
1.931	0.058	-0.089	5.470		
company_location_VN				-3.7090	4.790
-0.774	0.441	-13.263	5.845		
company_size_M				-1.5362	0.860
-1.785	0.079	-3.252	0.180		
company_size_S				-0.8028	0.962
-0.835	0.407	-2.721	1.116		
=====					
Omnibus:		24.077	Durbin-Watson:		2.067
Prob(Omnibus):		0.000	Jarque-Bera (JB):		67.549
Skew:		-0.519	Prob(JB):		2.15e-15
Kurtosis:		5.899	Cond. No.		5.87e+20
=====					

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The smallest eigenvalue is 9.49e-30. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

""

2.6 Define OSR2

```
[16]: # compute out-of-sample R-squared using the test set
def OSR2(model, df_train, df_test, dependent_var):
    y_test = df_test[dependent_var]
    y_pred = model.predict(df_test)
    SSE = np.sum((y_test - y_pred)**2)
    SST = np.sum((y_test - np.mean(df_train[dependent_var]))**2)
    return 1 - SSE/SST
```

```
[17]: OSR2(linreg, df_train, df_test, target_column)
```

```
[17]: 0.7394346635275322
```

2.7 Feature Selection: Use VIF to keep good features.

```
[18]: import statsmodels.api as sm
from statsmodels.stats.outliers_influence import variance_inflation_factor

def VIF(df, columns):
    values = sm.add_constant(df[columns]).values
    num_columns = len(columns) + 1
    vif = [variance_inflation_factor(values, i) for i in range(num_columns)]
    return pd.Series(vif[1:], index=columns)
```

2.8 Feature Selection: Identify and eliminate high P-Value features

```
[19]: def get_formula(features, target):
    features = [f for f in features if f != target]
    sum_features = ' + '.join(features)
    formula = '~ '.join([target, sum_features])
    return formula
```

```
[20]: def filter_feature(features, model, df_train, target):
    p_values = {}
    features = [f for f in model.params.index if f != 'Intercept']

    #
    for feat in features:
        p_values[feat] = model.pvalues.loc[feat]

    worst_feat = max(features, key=lambda f: p_values[f])

    print('WORST:', worst_feat, '-->', p_values[worst_feat])
    new_features = [f for f in features if f != worst_feat]

    new_model = smf.ols(formula = get_formula(new_features, target),
```

```
data = df_train).fit()
```

```
return new_features, new_model, p_values[worst_feat]
```

```
[21]: features = df_train.columns
model = linreg
p_value = float('inf')
models = {} # (formula, model)

while p_value > 0.05:
    features, model, p_value = filter_feature(features, model, df_train,
    ↪target_column)
    formula = get_formula(features, target_column)
```

```
WORST: company_location_CH --> 0.9917592810503727
WORST: job_title_Big_Data_Architect --> 0.9765611842923134
WORST: job_title_Data_Science_Consultant --> 0.978899374584294
WORST: company_location_PK --> 0.9751836220509351
WORST: company_location_CL --> 0.957280923019161
WORST: company_location_DE --> 0.9496630619778473
WORST: job_title_Computer_Vision_Engineer --> 0.937177139133903
WORST: work_year_2021e --> 0.9307019294832726
WORST: company_location_CA --> 0.9163775632788269
WORST: company_location_PL --> 0.9154613441099624
WORST: company_location_AT --> 0.9138669495937461
WORST: job_title_Research_Scientist --> 0.8906949325926343
WORST: job_title_Lead_Data_Analyst --> 0.9021539749130271
WORST: job_title_Data_Analyst --> 0.8980333768007385
WORST: company_location_PT --> 0.8149601176331617
WORST: company_location_HR --> 0.8150797190579797
WORST: job_title_Business_Data_Analyst --> 0.9202698235782161
WORST: employee_residence_HR --> 0.815079718972983
WORST: employee_residence_PH --> 0.9664697386595282
WORST: employee_residence_GR --> 0.8082642134004365
WORST: company_location_FR --> 0.6357524503392765
WORST: job_title_Cloud_Data_Engineer --> 0.9677198019038615
WORST: company_location_ES --> 0.7197205854084017
WORST: employee_residence_HK --> 0.9557818595661532
WORST: employee_residence_CL --> 0.9968502734836702
WORST: employee_residence_PL --> 0.6909557178946283
WORST: employee_residence_LU --> 0.9084730230282493
WORST: employee_residence_NZ --> 0.859307410598456
WORST: job_title_Head_of_Data_Science --> 0.7062949780221364
WORST: job_title_Big_Data_Engineer --> 0.7005089358665094
WORST: employee_residence_IR --> 0.6787798946154769
WORST: job_title_Staff_Data_Scientist --> 0.8792821168890983
```

WORST: company_location_HU --> 0.9272219478075725
WORST: company_location_IR --> 0.7744399997154151
WORST: employee_residence_RS --> 0.6787171506466998
WORST: company_location_IT --> 0.9016017008083217
WORST: job_title_Lead_Data_Engineer --> 0.6564565220889897
WORST: job_title_Computer_Vision_Software_Engineer --> 0.6686086550600603
WORST: employee_residence_DK --> 0.6523054428761657
WORST: job_title_Manager_Data_Science --> 0.6151040095322029
WORST: job_title_Data_Specialist --> 0.5535398310173966
WORST: employee_residence_NG --> 0.5415779583160978
WORST: company_location_LU --> 0.8670642552949288
WORST: company_location_NG --> 0.5415779580918841
WORST: employee_residence_CO --> 0.5750071339008482
WORST: company_location_CO --> 0.5750071338989861
WORST: job_title_Data_Analytics_Engineer --> 0.5316336596350422
WORST: job_title_ML_Engineer --> 0.519202249363028
WORST: job_title_Data_Analytics_Manager --> 0.5195121681344219
WORST: job_title_BI_Data_Analyst --> 0.5232920903632007
WORST: employee_residence_HU --> 0.5374865851441789
WORST: employee_residence_GB --> 0.5133020818990319
WORST: company_location_MT --> 0.5055915920116858
WORST: employee_residence_MT --> 0.5055915920465585
WORST: job_title_AI_Scientist --> 0.5075837760551528
WORST: remote_ratio --> 0.5354983173300677
WORST: employee_residence_CN --> 0.5378249025811064
WORST: job_title_Lead_Data_Scientist --> 0.4954750579896604
WORST: job_title_Data_Engineering_Manager --> 0.4632261450010775
WORST: job_title_Director_of_Data_Science --> 0.5523598612457511
WORST: job_title_Director_of_Data_Engineering --> 0.5021124501705805
WORST: job_title_Principal_Data_Analyst --> 0.4680701470325329
WORST: job_title_Machine_Learning_Infrastructure_Engineer --> 0.4856626869788869
WORST: employee_residence_SI --> 0.46050380187407713
WORST: company_location_SI --> 0.4605038018713522
WORST: job_title_Data_Scientist --> 0.4189258110639156
WORST: job_title_Marketing_Data_Analyst --> 0.40253054007965905
WORST: company_location_BR --> 0.41746030310376336
WORST: employee_residence_MD --> 0.3166879555856055
WORST: company_location_MD --> 0.31668795558148877
WORST: job_title_Data_Architect --> 0.3066123768719811
WORST: employee_residence_RO --> 0.317503351071737
WORST: job_title_Machine_Learning_Engineer --> 0.21901309459396162
WORST: employee_residence_BG --> 0.23769372183539975
WORST: job_title_Machine_Learning_Scientist --> 0.20449851004080027
WORST: job_title_Principal_Data_Scientist --> 0.23958043514479097
WORST: job_title_Data_Science_Engineer --> 0.19366650816381334
WORST: company_location_GR --> 0.1936665080268255
WORST: company_location_CN --> 0.18029210215867833
WORST: employee_residence_PT --> 0.20593951540626168

```

WORST: company_size_M --> 0.17824332034677423
WORST: job_title_Finance_Data_Analyst --> 0.17971681822916835
WORST: company_size_S --> 0.19610909727278517
WORST: company_location_RU --> 0.15365049646812592
WORST: job_title_Head_of_Data --> 0.15365049647602486
WORST: employment_type_FL --> 0.36007373222483796
WORST: company_location_NL --> 0.14438500198718263
WORST: job_title_Data_Engineer --> 0.21137667285849507
WORST: job_title_Applied_Data_Scientist --> 0.12074026356174956
WORST: employee_residence_PR --> 0.08765391158777444
WORST: employee_residence_VN --> 0.08750858729921741
WORST: employee_residence_AT --> 0.0711152222596231
WORST: company_location_DK --> 0.07944439646037656
WORST: employee_residence_IT --> 0.09345170753000334
WORST: employee_residence_ES --> 0.14387124916723143
WORST: employee_residence_FR --> 0.1827550794812623
WORST: job_title_Applied_Machine_Learning_Scientist --> 0.08398503678915212
WORST: job_title_Financial_Data_Analyst --> 0.06962268170459035
WORST: experience_level_MI --> 0.05951294621155277
WORST: company_location_AS --> 0.06033434807966886
WORST: employee_residence_RU --> 0.04207439881923185

```

```

[22]: best_features = features + ['employee_residence_RU']
      print(best_features)
      best_linreg = smf.ols(formula=get_formula(best_features, target_column),
                           data=df_train).fit()
      print(best_linreg.summary())

```

```

['salary_in_usd', 'experience_level_EX', 'experience_level_SE',
'employment_type_FT', 'employment_type_PT', 'job_title_Data_Science_Manager',
'job_title_Principal_Data_Engineer', 'job_title_Product_Data_Analyst',
'employee_residence_BE', 'employee_residence_BR', 'employee_residence_CA',
'employee_residence_DE', 'employee_residence_IN', 'employee_residence_JE',
'employee_residence_JP', 'employee_residence_KE', 'employee_residence_MX',
'employee_residence_NL', 'employee_residence_PK', 'employee_residence_SG',
'employee_residence_TR', 'employee_residence_UA', 'employee_residence_US',
'company_location_BE', 'company_location_GB', 'company_location_IL',
'company_location_IN', 'company_location_JP', 'company_location_KE',
'company_location_MX', 'company_location_NZ', 'company_location_SG',
'company_location_TR', 'company_location_UA', 'company_location_US',
'company_location_VN', 'employee_residence_RU']

```

OLS Regression Results

```

=====
Dep. Variable:          cbrt_salary_in_usd      R-squared:                0.966
Model:                  OLS                     Adj. R-squared:           0.959
Method:                 Least Squares            F-statistic:             143.4
Date:                  Mon, 08 May 2023          Prob (F-statistic):       3.01e-90
Time:                  01:04:35                  Log-Likelihood:          -391.77

```

No. Observations: 171 AIC: 841.5
Df Residuals: 142 BIC: 932.7
Df Model: 28
Covariance Type: nonrobust

		coef	std err	t	P> t
[0.025 0.975]					

Intercept		23.1892	1.824	12.712	0.000
19.583	26.795				
salary_in_usd		0.0001	3.52e-06	28.838	0.000
9.46e-05	0.000				
experience_level_EX		4.2508	1.177	3.612	0.000
1.925	6.577				
experience_level_SE		2.8723	0.496	5.792	0.000
1.892	3.853				
employment_type_FT		6.7035	1.745	3.842	0.000
3.254	10.153				
employment_type_PT		5.6287	2.302	2.445	0.016
1.078	10.180				
job_title_Data_Science_Manager		3.3110	1.286	2.575	0.011
0.769	5.853				
job_title_Principal_Data_Engineer		-16.0164	3.072	-5.214	0.000
-22.089	-9.944				
job_title_Product_Data_Analyst		-8.2223	2.746	-2.994	0.003
-13.651	-2.793				
employee_residence_BE		2.8731	1.336	2.150	0.033
0.232	5.515				
employee_residence_BR		-8.3231	1.649	-5.046	0.000
-11.584	-5.063				
employee_residence_CA		4.7788	1.408	3.394	0.001
1.996	7.562				
employee_residence_DE		3.7198	0.851	4.373	0.000
2.038	5.401				
employee_residence_IN		-9.6288	1.439	-6.692	0.000
-12.473	-6.784				
employee_residence_JE		6.3717	2.675	2.382	0.019
1.083	11.660				
employee_residence_JP		1.8942	0.711	2.664	0.009
0.489	3.299				
employee_residence_KE		-4.9128	1.333	-3.685	0.000
-7.548	-2.277				
employee_residence_MX		-4.9726	0.958	-5.189	0.000
-6.867	-3.078				
employee_residence_NL		3.9997	1.669	2.397	0.018
0.701	7.298				

employee_residence_PK	-9.9922	2.048	-4.878	0.000
-14.042 -5.943				
employee_residence_SG	3.6133	1.341	2.694	0.008
0.962 6.264				
employee_residence_TR	-2.3577	0.957	-2.465	0.015
-4.249 -0.467				
employee_residence_UA	-3.7505	1.333	-2.813	0.006
-6.386 -1.115				
employee_residence_US	3.1694	0.883	3.589	0.000
1.424 4.915				
company_location_BE	2.8731	1.336	2.150	0.033
0.232 5.515				
company_location_GB	3.3960	0.900	3.775	0.000
1.618 5.174				
company_location_IL	3.6133	1.341	2.694	0.008
0.962 6.264				
company_location_IN	5.5856	1.600	3.490	0.001
2.422 8.749				
company_location_JP	1.8942	0.711	2.664	0.009
0.489 3.299				
company_location_KE	-4.9128	1.333	-3.685	0.000
-7.548 -2.277				
company_location_MX	-4.9726	0.958	-5.189	0.000
-6.867 -3.078				
company_location_NZ	0	0	nan	nan
0 0				
company_location_SG	0	0	nan	nan
0 0				
company_location_TR	-2.3577	0.957	-2.465	0.015
-4.249 -0.467				
company_location_UA	-3.7505	1.333	-2.813	0.006
-6.386 -1.115				
company_location_US	2.1380	0.786	2.721	0.007
0.585 3.691				
company_location_VN	-14.4247	2.667	-5.408	0.000
-19.697 -9.152				
employee_residence_RU	4.3178	2.105	2.051	0.042
0.157 8.479				

Omnibus:	25.929	Durbin-Watson:	1.946
Prob(Omnibus):	0.000	Jarque-Bera (JB):	43.882
Skew:	-0.772	Prob(JB):	2.96e-10
Kurtosis:	4.943	Cond. No.	6.31e+21

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The smallest eigenvalue is 8.2e-32. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

```
[23]: OSR2(best_linreg, df_train, df_test, target_column)
```

```
[23]: 0.8035179130896895
```

2.9 Remove 0 coefficient and NaN p-value features

```
[24]: best_features.remove('company_location_NZ')
best_features.remove('company_location_SG')
best_linreg2 = smf.ols(formula=get_formula(best_features, target_column),
                      data=df_train).fit()
print(best_linreg2.summary())
```

```

                        OLS Regression Results
=====
Dep. Variable:          cbrt_salary_in_usd    R-squared:                0.966
Model:                  OLS                  Adj. R-squared:           0.959
Method:                 Least Squares        F-statistic:             143.4
Date:                  Mon, 08 May 2023      Prob (F-statistic):      3.01e-90
Time:                  01:04:36              Log-Likelihood:          -391.77
No. Observations:      171                  AIC:                    841.5
Df Residuals:          142                  BIC:                    932.7
Df Model:              28
Covariance Type:       nonrobust
=====
=====
```

	coef	std err	t	P> t
[0.025	0.975]			

Intercept	23.1892	1.824	12.712	0.000
19.583	26.795			
salary_in_usd	0.0001	3.52e-06	28.838	0.000
9.46e-05	0.000			
experience_level_EX	4.2508	1.177	3.612	0.000
1.925	6.577			
experience_level_SE	2.8723	0.496	5.792	0.000
1.892	3.853			
employment_type_FT	6.7035	1.745	3.842	0.000
3.254	10.153			
employment_type_PT	5.6287	2.302	2.445	0.016
1.078	10.180			
job_title_Data_Science_Manager	3.3110	1.286	2.575	0.011
0.769	5.853			
job_title_Principal_Data_Engineer	-16.0164	3.072	-5.214	0.000

-22.089	-9.944				
job_title_Product_Data_Analyst		-8.2223	2.746	-2.994	0.003
-13.651	-2.793				
employee_residence_BE		2.8731	1.336	2.150	0.033
0.232	5.515				
employee_residence_BR		-8.3231	1.649	-5.046	0.000
-11.584	-5.063				
employee_residence_CA		4.7788	1.408	3.394	0.001
1.996	7.562				
employee_residence_DE		3.7198	0.851	4.373	0.000
2.038	5.401				
employee_residence_IN		-9.6288	1.439	-6.692	0.000
-12.473	-6.784				
employee_residence_JE		6.3717	2.675	2.382	0.019
1.083	11.660				
employee_residence_JP		1.8942	0.711	2.664	0.009
0.489	3.299				
employee_residence_KE		-4.9128	1.333	-3.685	0.000
-7.548	-2.277				
employee_residence_MX		-4.9726	0.958	-5.189	0.000
-6.867	-3.078				
employee_residence_NL		3.9997	1.669	2.397	0.018
0.701	7.298				
employee_residence_PK		-9.9922	2.048	-4.878	0.000
-14.042	-5.943				
employee_residence_SG		3.6133	1.341	2.694	0.008
0.962	6.264				
employee_residence_TR		-2.3577	0.957	-2.465	0.015
-4.249	-0.467				
employee_residence_UA		-3.7505	1.333	-2.813	0.006
-6.386	-1.115				
employee_residence_US		3.1694	0.883	3.589	0.000
1.424	4.915				
company_location_BE		2.8731	1.336	2.150	0.033
0.232	5.515				
company_location_GB		3.3960	0.900	3.775	0.000
1.618	5.174				
company_location_IL		3.6133	1.341	2.694	0.008
0.962	6.264				
company_location_IN		5.5856	1.600	3.490	0.001
2.422	8.749				
company_location_JP		1.8942	0.711	2.664	0.009
0.489	3.299				
company_location_KE		-4.9128	1.333	-3.685	0.000
-7.548	-2.277				
company_location_MX		-4.9726	0.958	-5.189	0.000
-6.867	-3.078				
company_location_TR		-2.3577	0.957	-2.465	0.015

-4.249	-0.467				
company_location_UA		-3.7505	1.333	-2.813	0.006
-6.386	-1.115				
company_location_US		2.1380	0.786	2.721	0.007
0.585	3.691				
company_location_VN		-14.4247	2.667	-5.408	0.000
-19.697	-9.152				
employee_residence_RU		4.3178	2.105	2.051	0.042
0.157	8.479				

Omnibus:	25.929	Durbin-Watson:	1.946
Prob(Omnibus):	0.000	Jarque-Bera (JB):	43.882
Skew:	-0.772	Prob(JB):	2.96e-10
Kurtosis:	4.943	Cond. No.	4.06e+21

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The smallest eigenvalue is 1.98e-31. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

```
[25]: OSR2(best_linreg2, df_train, df_test, target_column)
```

```
[25]: 0.8035179130896896
```

2.10 Predict test set and compare to results.

```
[26]: y_pred = best_linreg2.predict(df_test)
      y_pred
```

```
[26]: 201    52.284403
      14    45.656034
      12    42.383180
      226   42.813622
      140   48.904506
      ...
      154   43.422710
      18    41.290900
      74    28.740143
      138   47.325491
      121   44.656114
      Length: 74, dtype: float64
```

```
[27]: df_test
```

```
[27]: salary_in_usd remote_ratio work_year_2021e experience_level_EX \
201      140000      100      1      0
14      103000      100      0      0
12      75966      100      1      0
226      75000      0      1      0
140      135000      100      0      0
..      ...      ...      ...      ...
154      81000      50      1      0
18      60000      100      1      0
74      28475      100      1      0
138      96357      50      1      0
121      93150      0      1      0
```

```
experience_level_MI experience_level_SE employment_type_FL \
201      0      1      0
14      1      0      0
12      1      0      0
226      1      0      0
140      1      0      0
..      ...      ...      ...
154      0      0      0
18      0      0      0
74      0      0      0
138      0      1      0
121      1      0      0
```

```
employment_type_FT employment_type_PT job_title_AI_Scientist ... \
201      1      0      0 ...
14      1      0      0 ...
12      1      0      0 ...
226      1      0      0 ...
140      1      0      0 ...
..      ...      ...      ...
154      1      0      0 ...
18      1      0      0 ...
74      1      0      0 ...
138      1      0      0 ...
121      1      0      0 ...
```

```
company_location_RU company_location_SG company_location_SI \
201      0      0      0
14      0      0      0
12      0      0      0
226      0      0      0
140      0      0      0
..      ...      ...      ...
154      0      0      0
```

18	0	0	0
74	0	0	0
138	0	0	0
121	0	0	0

	company_location_TR	company_location_UA	company_location_US	\
201	0	0	1	
14	0	0	1	
12	0	0	0	
226	0	0	1	
140	0	0	1	
..	
154	0	0	1	
18	0	0	1	
74	0	0	0	
138	0	0	0	
121	0	0	1	

	company_location_VN	company_size_M	company_size_S	cbirt_salary_in_usd
201	0	0	0	51.924941
14	0	0	0	46.875481
12	0	0	0	42.351918
226	0	0	0	42.171633
140	0	0	0	51.299278
..
154	0	0	1	43.267487
18	0	0	1	39.148676
74	0	1	0	30.536640
138	0	0	0	45.845258
121	0	1	0	45.330894

[74 rows x 138 columns]

2.11 Predict my own entries (randomized).

2.11.1 Read CSV again with new dataframe, dropping unnecessary features.

```
[28]: df_copy = pd.read_csv(r"Data Science Jobs Salaries.csv").
      ↪drop(columns=['salary_currency', 'salary_in_usd', 'salary'])
df_copy
```

```
[28]: work_year experience_level employment_type job_title \
0      2021e          EN          FT  Data Science Consultant
1      2020          SE          FT      Data Scientist
2      2021e          EX          FT  Head of Data Science
3      2021e          EX          FT      Head of Data
4      2021e          EN          FT  Machine Learning Engineer
```

..
240	2020	SE	FT	Data Scientist
241	2021e	MI	FT	Principal Data Scientist
242	2020	EN	FT	Data Scientist
243	2020	EN	CT	Business Data Analyst
244	2021e	SE	FT	Data Science Manager

	employee_residence	remote_ratio	company_location	company_size
0	DE	50	DE	L
1	GR	100	US	L
2	RU	0	RU	M
3	RU	50	RU	L
4	US	100	US	S
..
240	US	100	US	L
241	US	100	US	L
242	US	100	US	S
243	US	100	US	L
244	IN	50	IN	L

[245 rows x 8 columns]

2.11.2 Get unique values of each of our features to randomize.

```
[29]: possible_WY = df_copy['work_year'].unique()
possible_ET = df_copy['employment_type'].unique()
possible_ER = df_copy['employee_residence'].unique()
possible_RR = df_copy['remote_ratio'].unique()
possible_CL = df_copy['company_location'].unique()
possible_CS = df_copy['company_size'].unique()
```

2.11.3 Make an empty dataframe to add our imaginary data scientists to.

```
[30]: random_df_copy = df_copy.copy()
random_df_copy = random_df_copy[0:0]
random_df_copy
```

```
[30]: Empty DataFrame
Columns: [work_year, experience_level, employment_type, job_title,
employee_residence, remote_ratio, company_location, company_size]
Index: []
```

2.11.4 Create 200 imaginary data scientists by randomizing possible values from our raw dataset.

```
[31]: for i in range(200):

    EN_JT_index = random.randint(0, len(df_copy.index)-1)

    random_WY = random.choice(possible_WY)
    random_EL = df_copy.iloc[EN_JT_index]['experience_level']
    random_ET = random.choice(possible_ET)
    random_JT = df_copy.iloc[EN_JT_index]['job_title']
    random_ER = random.choice(possible_ER)
    random_RR = random.choice(possible_RR)
    random_CL = random.choice(possible_CL)
    random_CS = random.choice(possible_CS)

    random_df_copy.loc[len(random_df_copy.index)] = [random_WY, random_EL,
    random_ET, random_JT,
    random_ER, random_RR,
    random_CL, random_CS]

random_df_copy
```

```
[31]: work_year experience_level employment_type job_title \
0      2021e          EN          FL      Data Scientist
1      2021e          SE          FT  Machine Learning Scientist
2      2020          EN          CT    Data Science Consultant
3      2020          MI          PT      Data Engineer
4      2020          MI          PT      Data Analyst
..      ...          ...          ...          ...
195     2021e          SE          FL    Data Science Manager
196     2020          SE          CT  Data Engineering Manager
197     2020          SE          PT      Data Scientist
198     2020          EN          CT      Data Scientist
199     2021e          EN          FL      Data Scientist

employee_residence remote_ratio company_location company_size
0                UA           50             SI           L
1                US          100             ES           L
2                TR           0              CN           L
3                JP           0              IR           M
4                SG           0              ES           M
..              ...          ...          ...          ...
195             GR           50             IL           L
196             MT           50             RU           L
197             US           50             CH           M
198             DK          100             NZ           S
```

199	JP	50	PK	S
-----	----	----	----	---

[200 rows x 8 columns]

2.11.5 Create dummy variables for our features to use in making predictions.

```
[32]: random_df_copy = pd.get_dummies(random_df_copy)
      random_df_copy
```

```
[32]:
```

	remote_ratio	work_year_2020	work_year_2021e	experience_level_EN	\
0	50	0	1	1	
1	100	0	1	0	
2	0	1	0	1	
3	0	1	0	0	
4	0	1	0	0	
..	
195	50	0	1	0	
196	50	1	0	0	
197	50	1	0	0	
198	100	1	0	1	
199	50	0	1	1	

	experience_level_EX	experience_level_MI	experience_level_SE	\
0	0	0	0	
1	0	0	1	
2	0	0	0	
3	0	1	0	
4	0	1	0	
..	
195	0	0	1	
196	0	0	1	
197	0	0	1	
198	0	0	0	
199	0	0	0	

	employment_type_CT	employment_type_FL	employment_type_FT	...	\
0	0	1	0	...	
1	0	0	1	...	
2	1	0	0	...	
3	0	0	0	...	
4	0	0	0	...	
..	
195	0	1	0	...	
196	1	0	0	...	
197	0	0	0	...	
198	1	0	0	...	
199	0	1	0	...	

	company_location_RU	company_location_SG	company_location_SI	\
0	0	0	1	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	
..	
195	0	0	0	
196	1	0	0	
197	0	0	0	
198	0	0	0	
199	0	0	0	

	company_location_TR	company_location-UA	company_location_US	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	
..	
195	0	0	0	
196	0	0	0	
197	0	0	0	
198	0	0	0	
199	0	0	0	

	company_location_VN	company_size_L	company_size_M	company_size_S
0	0	1	0	0
1	0	1	0	0
2	0	1	0	0
3	0	0	1	0
4	0	0	1	0
..
195	0	1	0	0
196	0	1	0	0
197	0	0	1	0
198	0	0	0	1
199	0	0	0	1

[200 rows x 138 columns]

2.11.6 Rename columns to include `_` in existing ones with spaces (to match the features in the model).

```
[33]: for col in random_df_copy.columns:
        random_df_copy.rename(columns={col : '_' + col.split()[1]}, inplace=True)
```

2.11.7 Add 0 columns for features which weren't randomly chosen from possible values. If this happens, the dummy variables for those possible categories are not in our testing set and cannot work with the model as intended.

```
[34]: for c in df.columns:
        if c not in random_df_copy.columns:
            random_df_copy[c] = 0
```

2.11.8 Remove unneeded columns to match model perfectly.

```
[35]: to_drop = ['work_year_2020', 'experience_level_EN', 'employment_type_CT',
                ↪ 'job_title_3D_Computer_Vision_Researcher',
                'employee_residence_AE', 'company_location_AE', 'company_size_L',
                ↪ 'log_salary_in_usd', 'sqrt_salary_in_usd', 'cbrt_salary_in_usd']
        for c in to_drop:
            if c in random_df_copy.columns:
                random_df_copy.drop(columns=[c], inplace=True)
```

2.11.9 Print the processed, randomized dataframe of 200 data scientists.

```
[36]: random_df_copy
```

```
[36]:
```

	remote_ratio	work_year_2021e	experience_level_EX	experience_level_MI	\
0	50	1	0	0	
1	100	1	0	0	
2	0	0	0	0	
3	0	0	0	1	
4	0	0	0	1	
..	
195	50	1	0	0	
196	50	0	0	0	
197	50	0	0	0	
198	100	0	0	0	
199	50	1	0	0	

	experience_level_SE	employment_type_FL	employment_type_FT	\
0	0	1	0	
1	1	0	1	
2	0	0	0	
3	0	0	0	

4	0	0	0
..
195	1	1	0
196	1	0	0
197	1	0	0
198	0	0	0
199	0	1	0

	employment_type_PT	job_title_AI_Scientist	\
0	0	0	
1	0	0	
2	0	0	
3	1	0	
4	1	0	
..	
195	0	0	
196	0	0	
197	1	0	
198	0	0	
199	0	0	

	job_title_Applied_Data_Scientist	...	company_location_US	\
0	0	...	0	
1	0	...	0	
2	0	...	0	
3	0	...	0	
4	0	...	0	
..	
195	0	...	0	
196	0	...	0	
197	0	...	0	
198	0	...	0	
199	0	...	0	

	company_location_VN	company_size_M	company_size_S	salary_in_usd	\
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	1	0	0	
4	0	1	0	0	
..	
195	0	0	0	0	
196	0	0	0	0	
197	0	1	0	0	
198	0	0	1	0	
199	0	0	1	0	

	job_title_Big_Data_Architect \
0	0
1	0
2	0
3	0
4	0
..	...
195	0
196	0
197	0
198	0
199	0

	job_title_Machine_Learning_Infrastructure_Engineer \
0	0
1	0
2	0
3	0
4	0
..	...
195	0
196	0
197	0
198	0
199	0

	job_title_Marketing_Data_Analyst	job_title_Principal_Data_Engineer \
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
..
195	0	0
196	0	0
197	0	0
198	0	0
199	0	0

	job_title_Staff_Data_Scientist
0	0
1	0
2	0
3	0
4	0
..	...
195	0

```

196          0
197          0
198          0
199          0

```

```
[200 rows x 137 columns]
```

2.11.10 Make predictions for our 200 random data scientist's salaries.

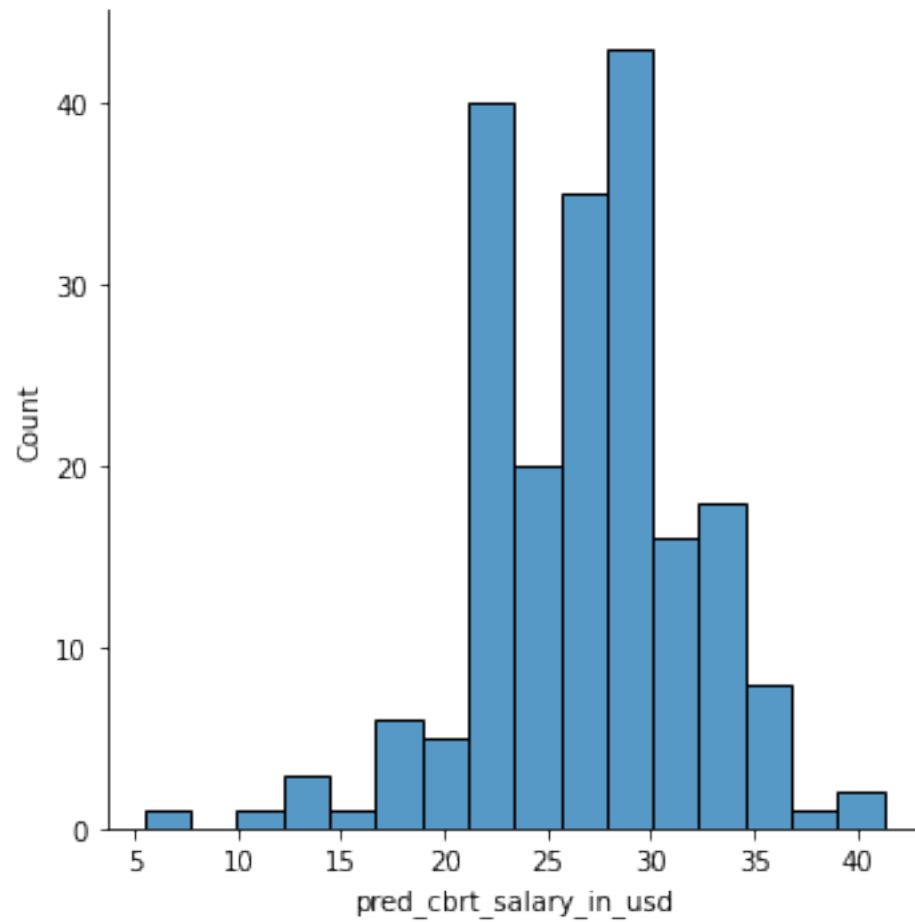
```
[37]: y_pred_rand = best_linreg2.predict(random_df_copy)
      # y_pred_rand = np rint(y_pred_rand).astype(int)
      y_pred_rand
```

```

[37]: 0      19.438706
      1      35.934362
      2      20.831496
      3      30.711979
      4      32.431146
      ...
      195     32.985860
      196     26.061482
      197     34.859494
      198     23.189158
      199     25.083323
      Length: 200, dtype: float64

```

```
[38]: random_df_copy['pred_cbrt_salary_in_usd'] = y_pred_rand
      sns.displot(data=random_df_copy, x = 'pred_cbrt_salary_in_usd')
      plt.show();
```



```
[39]: negative_predictions = random_df_copy[random_df_copy['pred_cbrt_salary_in_usd']  
      ↪ < 0]  
      display(negative_predictions)
```

Empty DataFrame

```

Columns: [remote_ratio, work_year_2021e, experience_level_EX,
↳experience_level_MI, experience_level_SE, employment_type_FL,
↳employment_type_FT, employment_type_PT, job_title_AI_Scientist,
↳job_title_Applied_Data_Scientist,
↳job_title_Applied_Machine_Learning_Scientist, job_title_BI_Data_Analyst,
↳job_title_Big_Data_Engineer, job_title_Business_Data_Analyst,
↳job_title_Cloud_Data_Engineer, job_title_Computer_Vision_Engineer,
↳job_title_Computer_Vision_Software_Engineer, job_title_Data_Analyst,
↳job_title_Data_Analytics_Engineer, job_title_Data_Analytics_Manager,
↳job_title_Data_Architect, job_title_Data_Engineer,
↳job_title_Data_Engineering_Manager, job_title_Data_Science_Consultant,
↳job_title_Data_Science_Engineer, job_title_Data_Science_Manager,
↳job_title_Data_Scientist, job_title_Data_Specialist,
↳job_title_Director_of_Data_Engineering, job_title_Director_of_Data_Science,
↳job_title_Finance_Data_Analyst, job_title_Financial_Data_Analyst,
↳job_title_Head_of_Data, job_title_Head_of_Data_Science,
↳job_title_Lead_Data_Analyst, job_title_Lead_Data_Engineer,
↳job_title_Lead_Data_Scientist, job_title_ML_Engineer,
↳job_title_Machine_Learning_Engineer, job_title_Machine_Learning_Scientist,
↳job_title_Manager_Data_Science, job_title_Principal_Data_Analyst,
↳job_title_Principal_Data_Scientist, job_title_Product_Data_Analyst,
↳job_title_Research_Scientist, employee_residence_AT, employee_residence_BE,
↳employee_residence_BG, employee_residence_BR, employee_residence_CA,
↳employee_residence_CL, employee_residence_CN, employee_residence_CO,
↳employee_residence_DE, employee_residence_DK, employee_residence_ES,
↳employee_residence_FR, employee_residence_GB, employee_residence_GR,
↳employee_residence_HK, employee_residence_HR, employee_residence_HU,
↳employee_residence_IN, employee_residence_IR, employee_residence_IT,
↳employee_residence_JE, employee_residence_JP, employee_residence_KE,
↳employee_residence_LU, employee_residence_MD, employee_residence_MT,
↳employee_residence_MX, employee_residence_NG, employee_residence_NL,
↳employee_residence_NZ, employee_residence_PH, employee_residence_PK,
↳employee_residence_PL, employee_residence_PR, employee_residence_PT,
↳employee_residence_RO, employee_residence_RS, employee_residence_RU,
↳employee_residence_SG, employee_residence_SI, employee_residence_TR,
↳employee_residence_UA, employee_residence_US, employee_residence_VN,
↳company_location_AS, company_location_AT, company_location_BE,
↳company_location_BR, company_location_CA, company_location_CH,
↳company_location_CL, company_location_CN, company_location_CO,
↳company_location_DE, company_location_DK, ...]

```

```
Index: []
```

```
[0 rows x 138 columns]
```

```

[41]: x_min = np.min(y_pred)
      x_max = np.max(y_pred)
      y_min = np.min(df_test_y_actual)

```

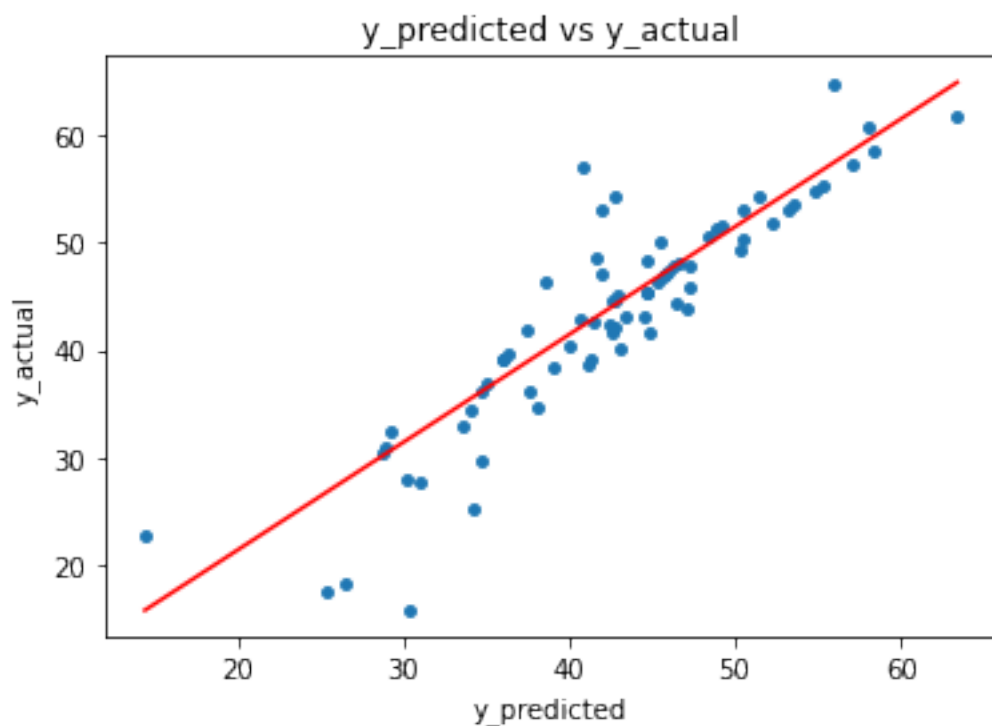
```

y_max = np.max(df_test_y_actual)

x = np.arange(x_min, x_max + 1)
y = np.arange(y_min, y_max + 1)

plt.scatter(x=y_pred, y=df_test_y_actual, s=15)
plt.plot(x, y, color='red') # y = x
plt.xlabel('y_predicted')
plt.ylabel('y_actual')
plt.title('y_predicted vs y_actual')
plt.show();

```



```

[42]: residuals = df_test_y_actual - y_pred

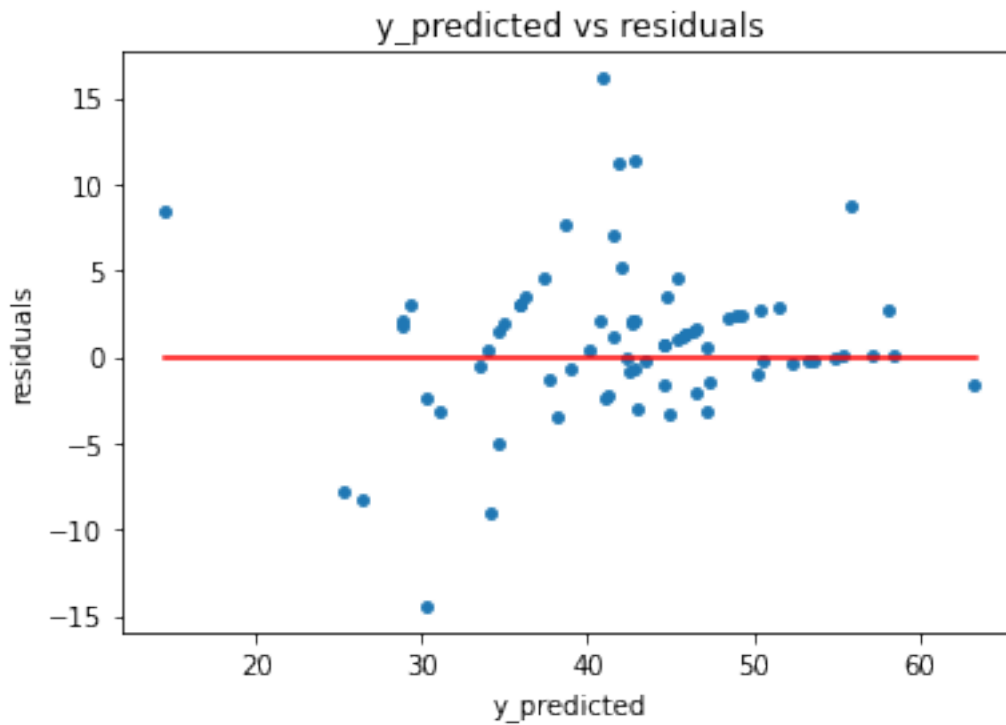
x_min = np.min(y_pred)
x_max = np.max(y_pred)
y_min = np.min(residuals)
y_max = np.max(residuals)

x = np.arange(x_min, x_max + 1)
y = np.zeros(len(x))

```



```
plt.scatter(x=y_pred, y=residuals, s=15)
plt.plot(x, y, color='red') #  $y = x$ 
plt.xlabel('y_predicted')
plt.ylabel('residuals')
plt.title('y_predicted vs residuals')
plt.show();
```



```
[43]: error = df_test_y_actual - y_pred
```