

```
In [1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
%matplotlib inline
```

```
In [2]: divorce = pd.read_csv("divorce.csv")
divorce.dtypes
```

```
Out[2]: divorce_date      object
dob_man      object
education_man object
income_man    float64
dob_woman     object
education_woman object
income_woman  float64
marriage_date object
marriage_duration float64
num_kids      float64
dtype: object
```

```
In [3]: divorce = pd.read_csv("divorce.csv" , parse_dates=["marriage_date"])
divorce.dtypes
```

```
Out[3]: divorce_date      object
dob_man      object
education_man object
income_man    float64
dob_woman     object
education_woman object
income_woman  float64
marriage_date    datetime64[ns]
marriage_duration float64
num_kids      float64
dtype: object
```

```
In [4]: divorce["marriage_date"] = pd.to_datetime(divorce["marriage_date"])
divorce.dtypes
```

```
Out[4]: divorce_date      object
dob_man      object
education_man object
income_man    float64
dob_woman     object
education_woman object
income_woman  float64
marriage_date    datetime64[ns]
marriage_duration float64
num_kids      float64
dtype: object
```

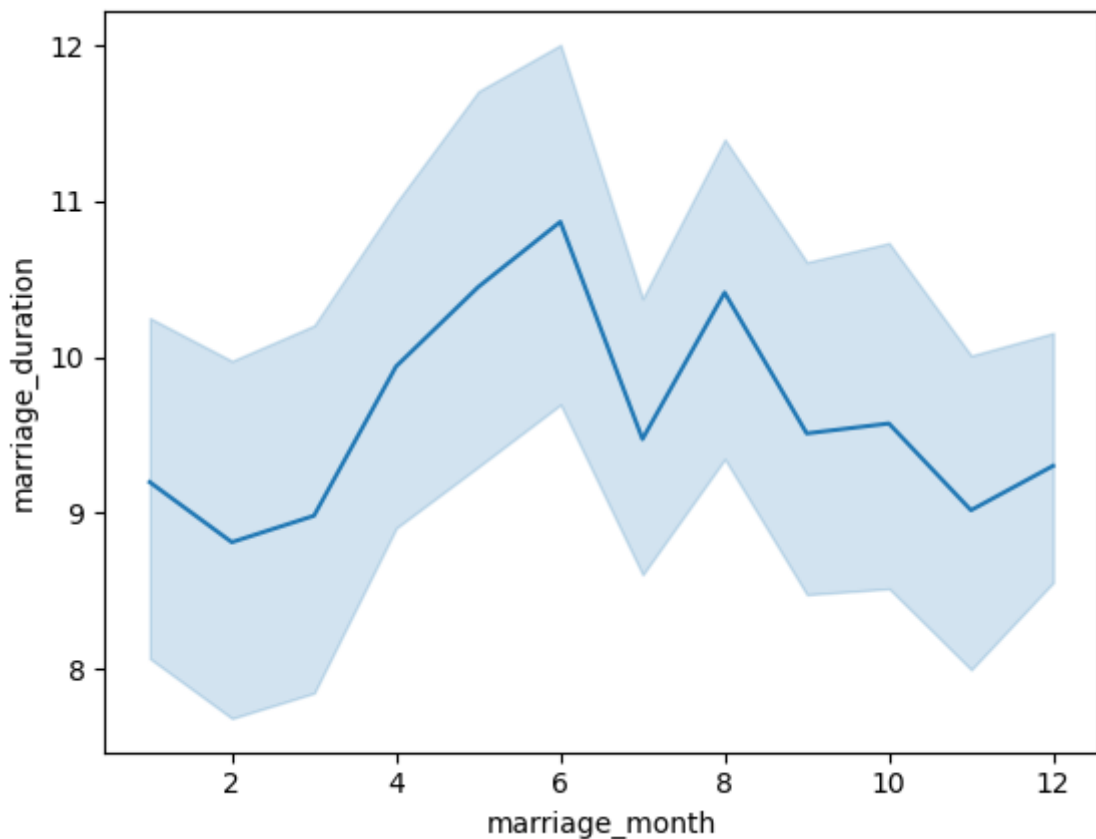
```
In [5]: # divorce["marriage_date"] = pd.to_datetime(divorce[["month" , "day" , "year"]])
# divorce.head(2)
```

```
In [6]: divorce["marriage_month"] = divorce["marriage_date"].dt.month
divorce.head()
```

Out[6]:

	divorce_date	dob_man	education_man	income_man	dob_woman	education_woman	inc
0	2006-09-06	1975-12-18	Secondary	2000.0	1983-08-01	Secondary	
1	2008-01-02	1976-11-17	Professional	6000.0	1977-03-13	Professional	
2	2011-01-02	1969-04-06	Preparatory	5000.0	1970-02-16	Professional	
3	2011-01-02	1979-11-13	Secondary	12000.0	1981-05-13	Secondary	
4	2011-01-02	1982-09-20	Professional	6000.0	1988-01-30	Professional	

In [7]: `sns.lineplot(data=divorce, x="marriage_month", y="marriage_duration")`  
`plt.show()`



In [8]: `DataFrame = pd.read_csv("divorce.csv")`  
`DataFrame.dtypes`

Out[8]:

divorce_date	object
dob_man	object
education_man	object
income_man	float64
dob_woman	object
education_woman	object
income_woman	float64
marriage_date	object
marriage_duration	float64
num_kids	float64
dtype:	object

1

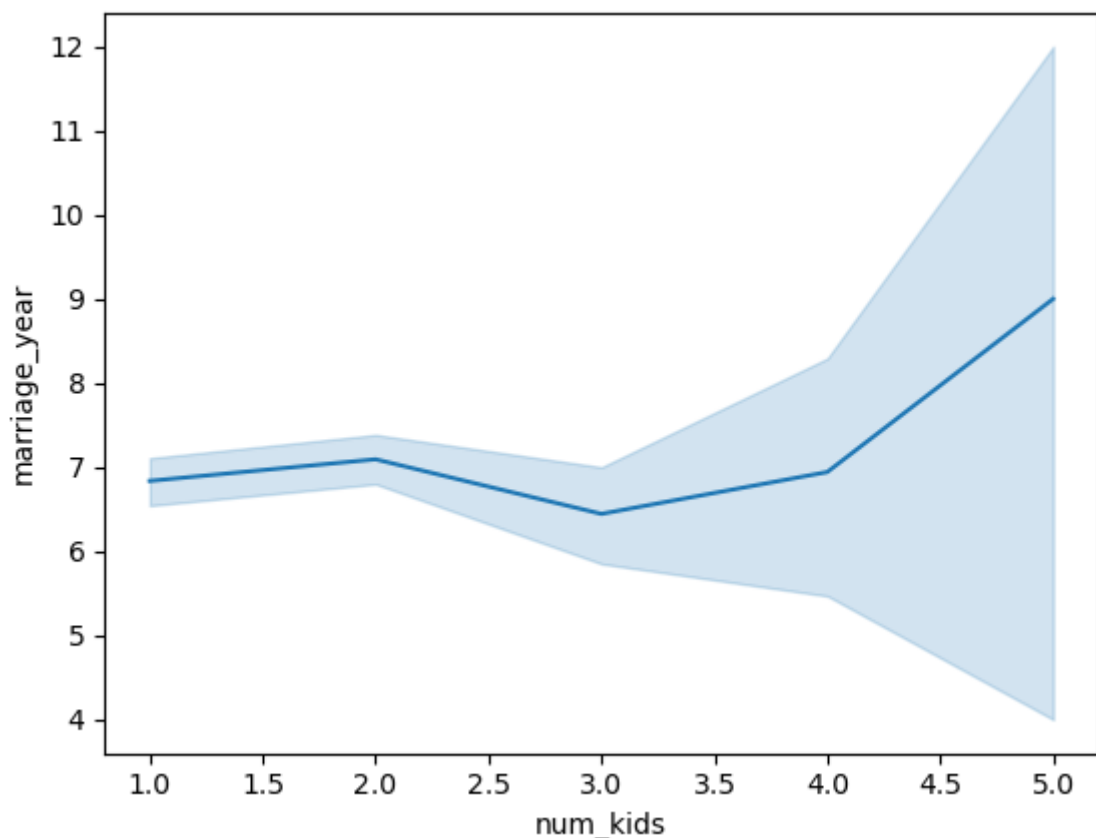
```
In [10]: divorce = pd.read_csv("divorce.csv" , parse_dates=["divorce_date" , "dob_man", "dob_woman"])
divorce.dtypes
```

```
Out[10]: divorce_date      datetime64[ns]
dob_man      datetime64[ns]
education_man      object
income_man      float64
dob_woman      datetime64[ns]
education_woman      object
income_woman      float64
marriage_date      object
marriage_duration      float64
num_kids      float64
dtype: object
```

2

```
In [11]: divorce["marriage_date"] = pd.to_datetime(divorce["marriage_date"])
divorce["marriage_year"] = divorce["marriage_date"].dt.month
sns.lineplot(data=divorce, x = "num_kids" , y = "marriage_year")
```

```
Out[11]: <AxesSubplot:xlabel='num_kids', ylabel='marriage_year'>
```



## CORRELATION

```
In [47]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
import numpy as np
%matplotlib inline
```

```
In [48]: divorce = pd.read_csv("divorce.csv")
divorce.corr()
```

```
Out[48]:
```

	income_man	income_woman	marriage_duration	num_kids
income_man	1.000000	0.318047	0.085321	0.040848
income_woman	0.318047	1.000000	0.078677	-0.018015
marriage_duration	0.085321	0.078677	1.000000	0.447358
num_kids	0.040848	-0.018015	0.447358	1.000000

## HEATMAP

```
In [49]: sns.heatmap(divorce.corr(), annot = True)
plt.show()
```



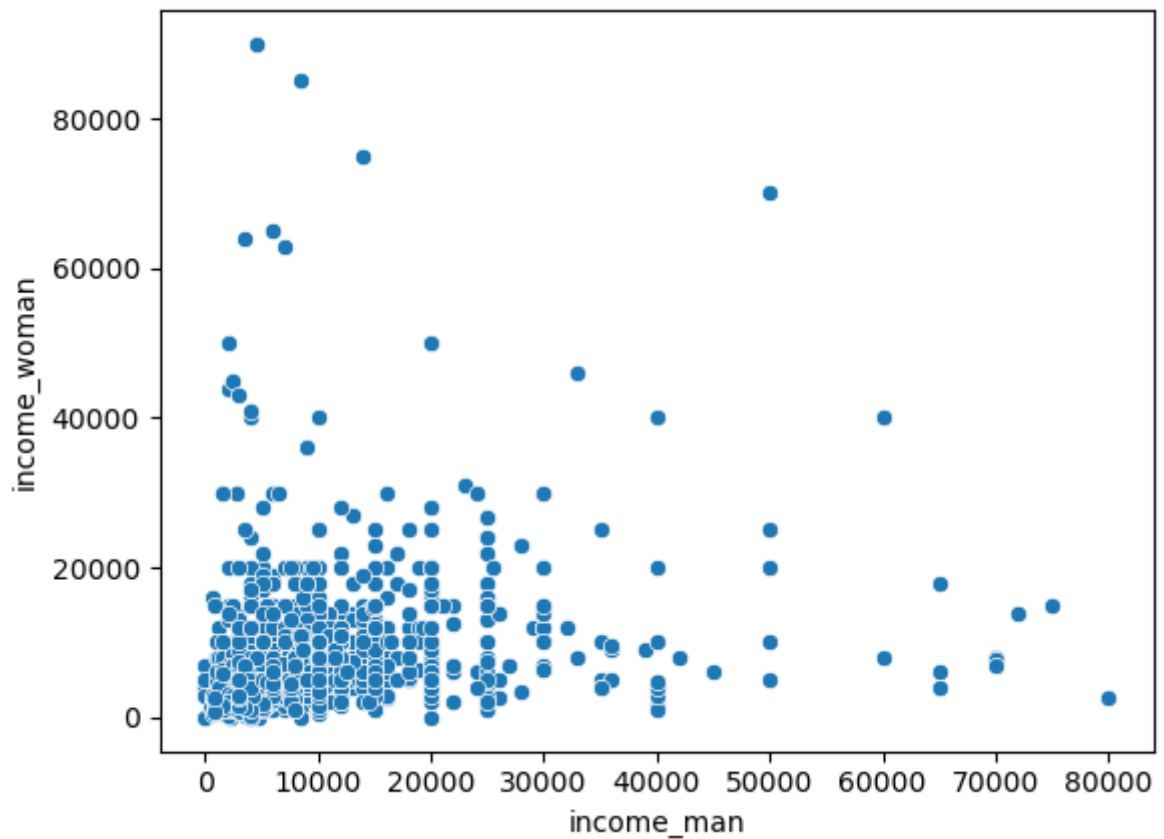
```
In [50]: divorce["divorce_date"].min()
```

```
Out[50]: '2000-01-08'
```

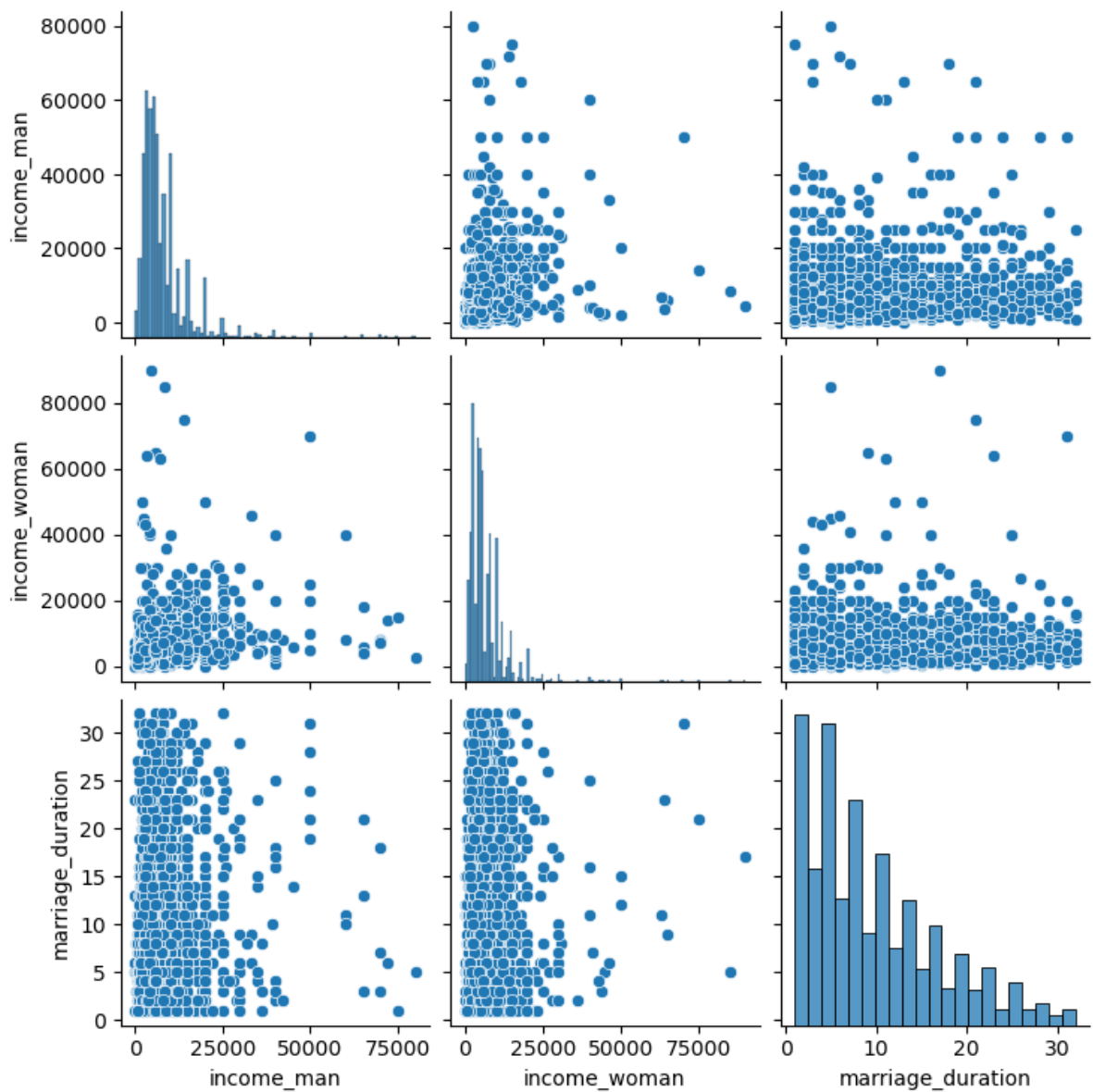
```
In [51]: divorce["divorce_date"].max()
```

```
Out[51]: '2015-11-03'
```

```
In [52]: sns.scatterplot(data=divorce, x="income_man", y="income_woman")  
plt.show()
```

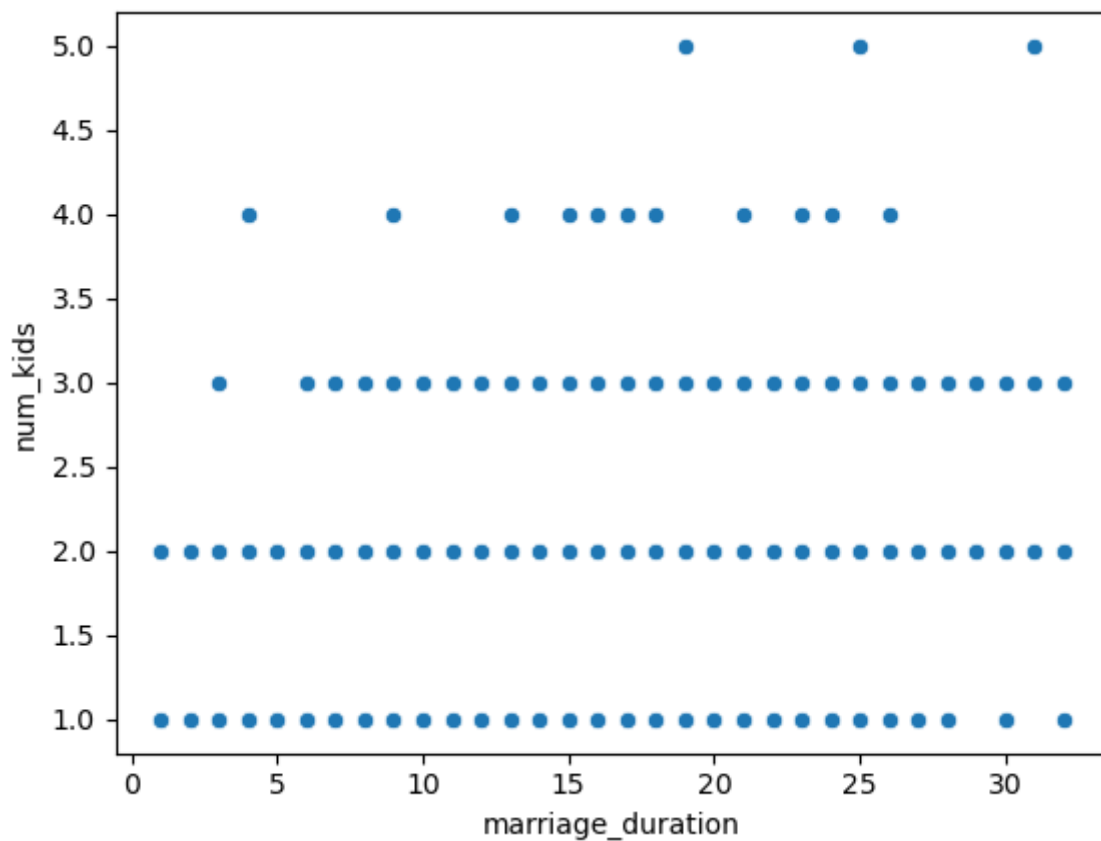


```
In [53]: sns.pairplot(data=divorce, vars=["income_man", "income_woman", "marriage_d  
plt.show()
```



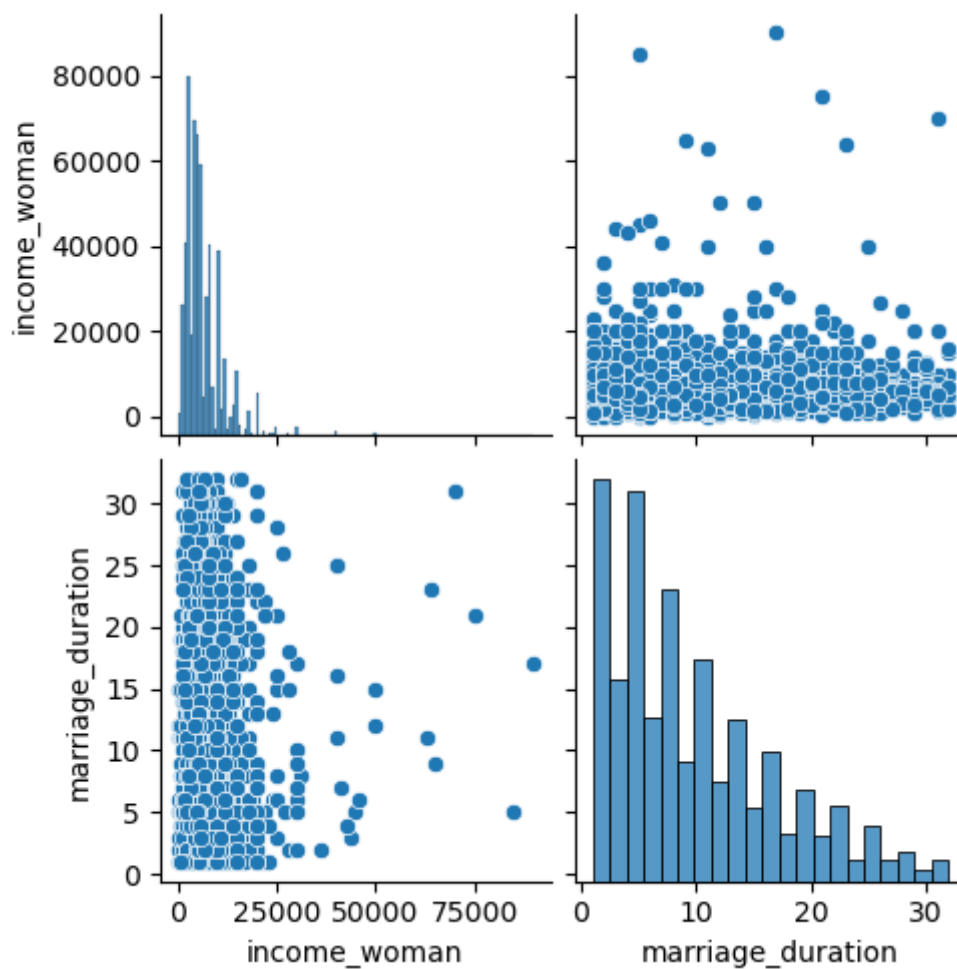
### 3

```
In [54]: sns.scatterplot(data =divorce, x = "marriage_duration" , y= "num_kids")  
plt.show()
```



## 4

```
In [55]: sns.pairplot(data=divorce , vars=["income_woman" , "marriage_duration"])
plt.show()
```

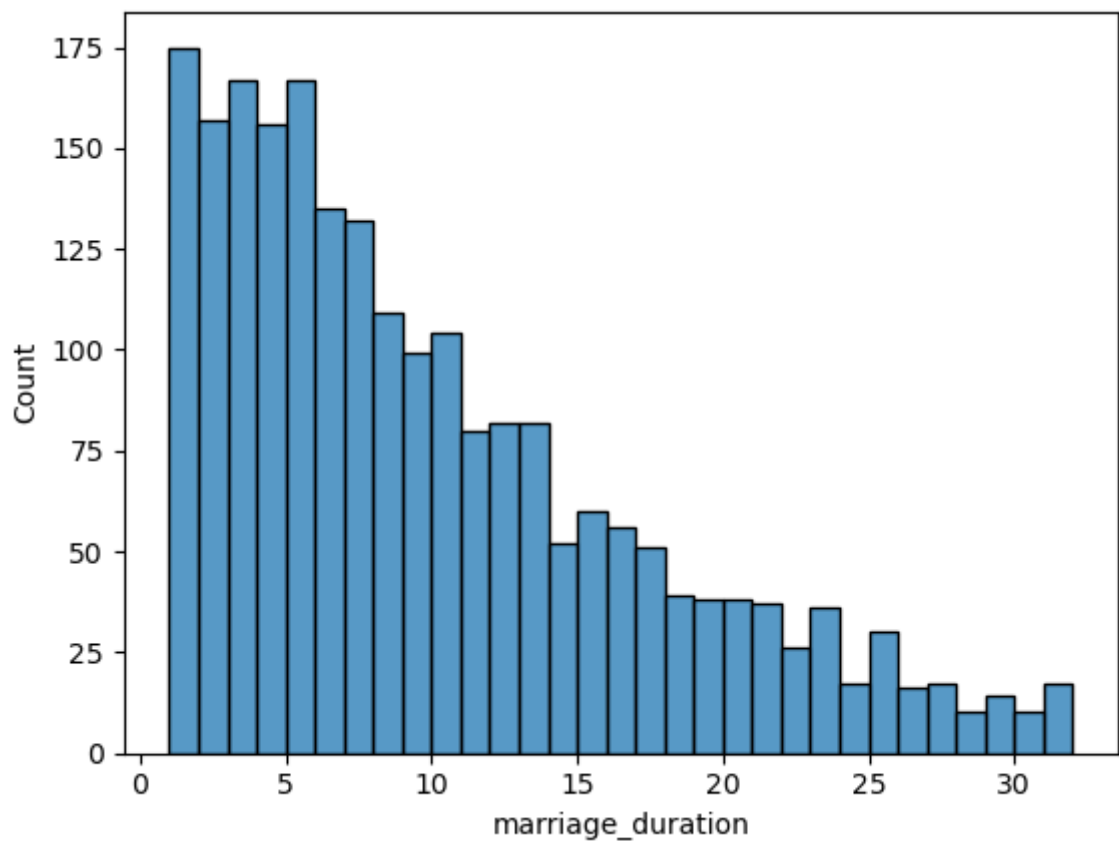


```
In [56]: divorce["education_man"].value_counts()
```

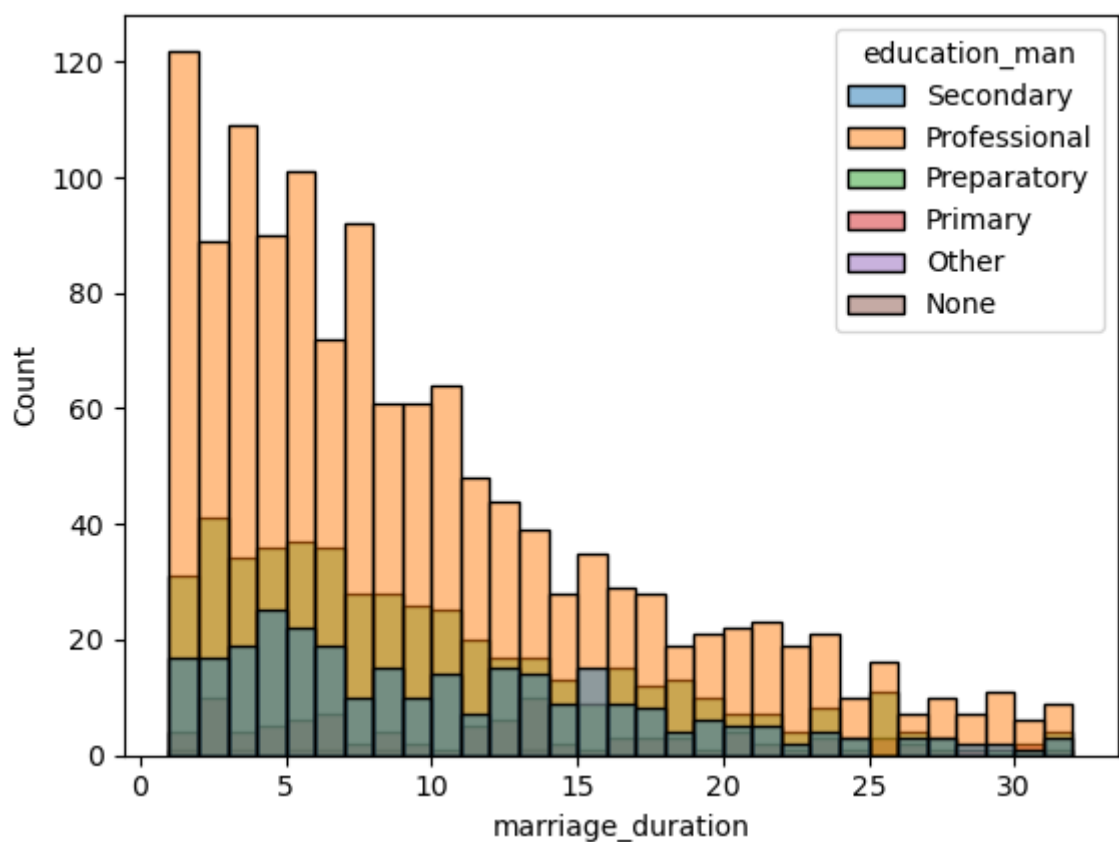
```
Out[56]: Professional    1313
Preparatory             501
Secondary                288
Primary                 100
None                     4
Other                    3
Name: education_man, dtype: int64
```

```
In [57]: sns.histplot(data=divorce, x = "marriage_duration" , binwidth= 1)
plt.show()
```

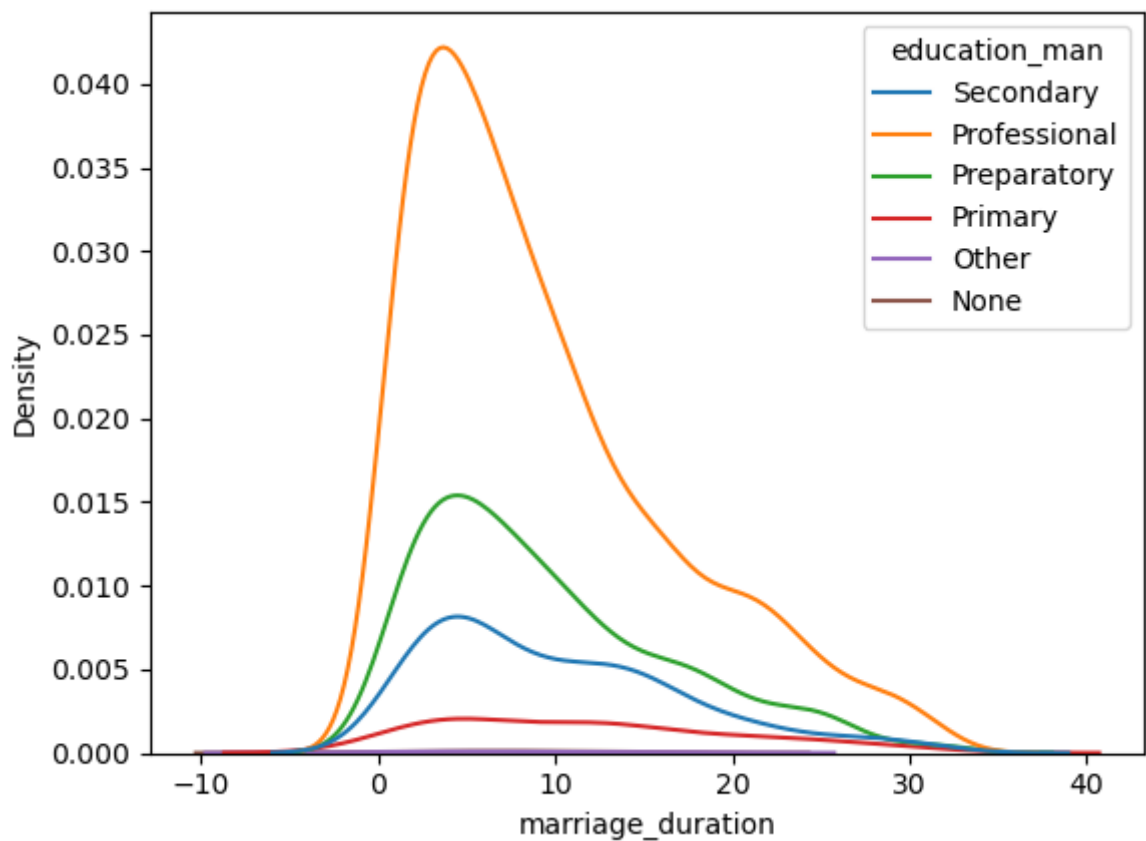




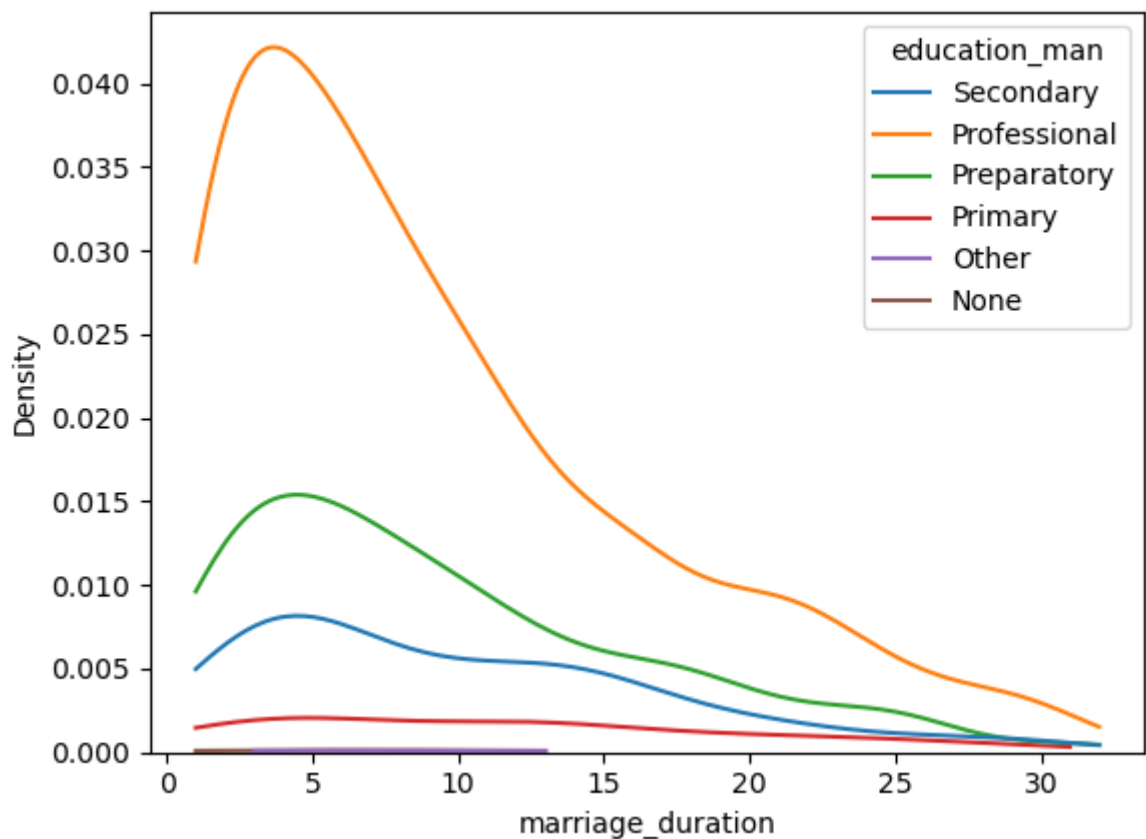
```
In [58]: sns.histplot(data = divorce , x = "marriage_duration" , hue= "education_man"  
plt.show())
```



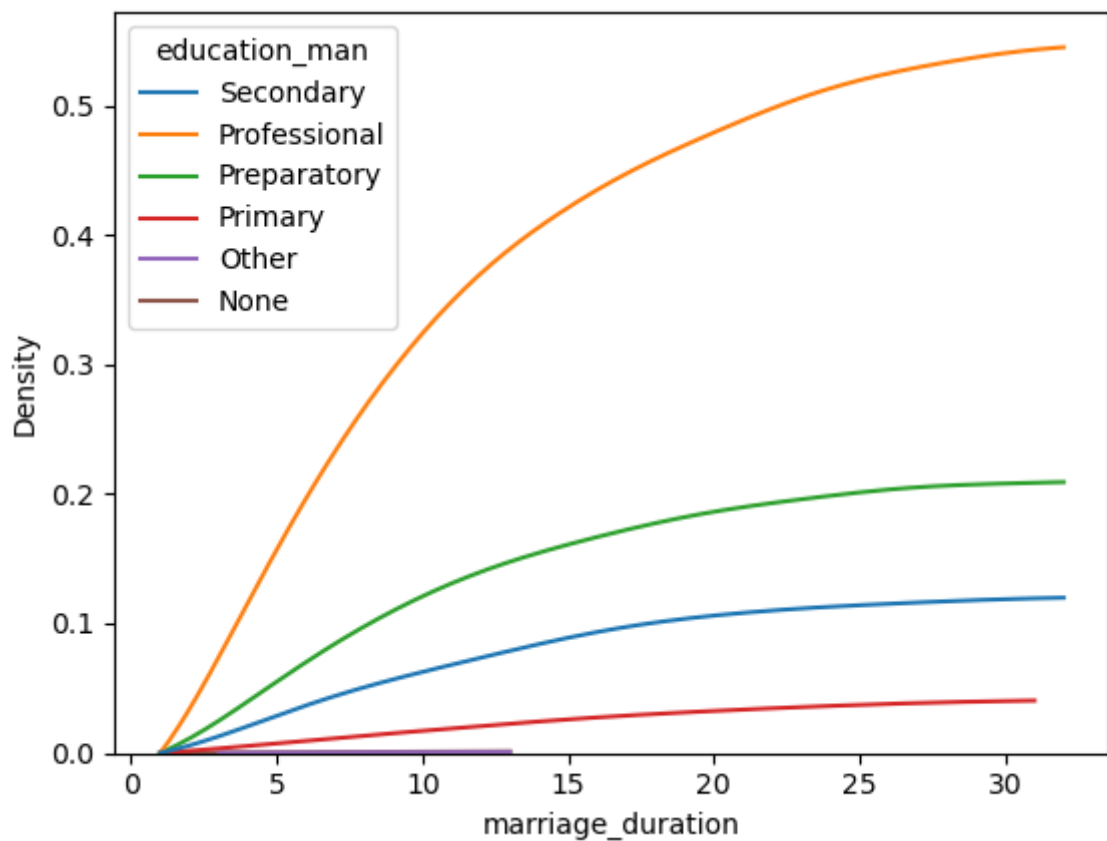
```
In [59]: sns.kdeplot(data= divorce , x = "marriage_duration" , hue = "education_man")  
plt.show())
```



```
In [60]: sns.kdeplot(data=divorce , x = "marriage_duration" , hue="education_man" , cu  
plt.show()
```



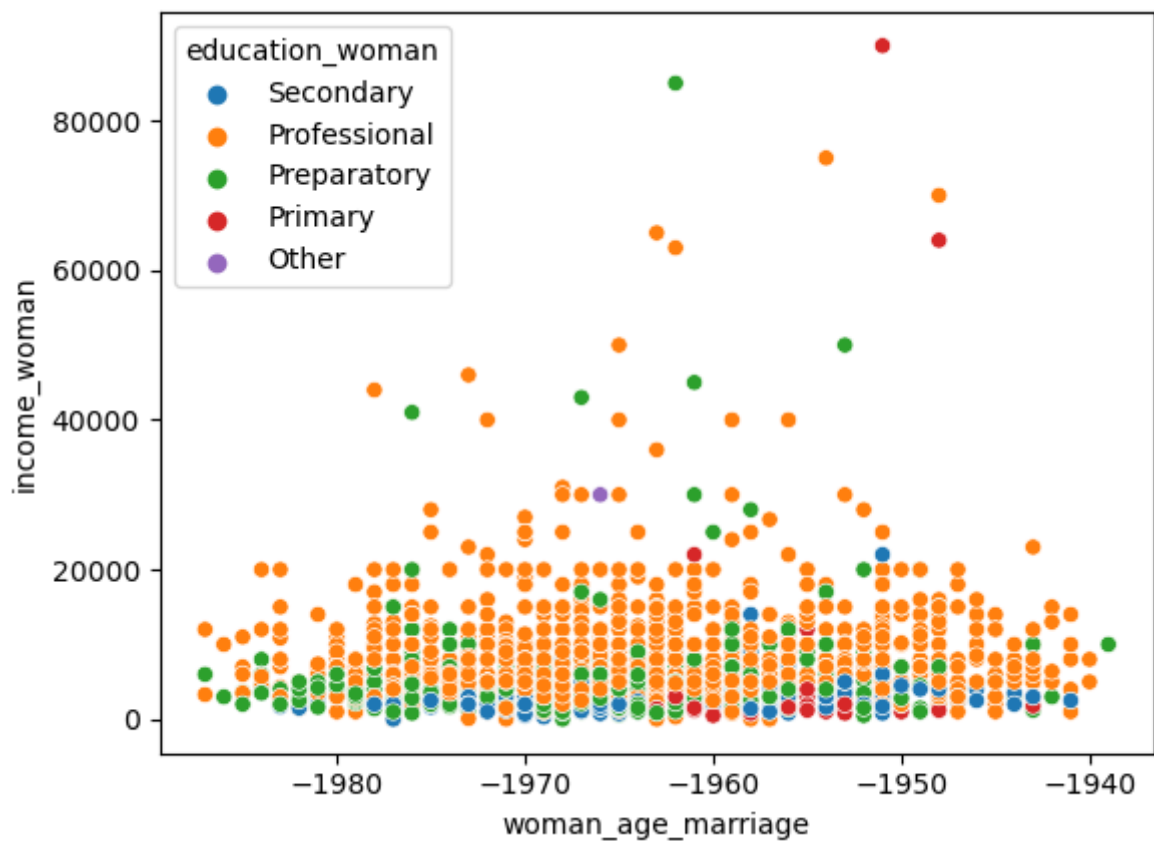
```
In [61]: sns.kdeplot(data = divorce , x = "marriage_duration" , hue="education_man" ,  
plt.show()
```



```
In [12]: divorce["man_age_marriage"] = divorce["marriage_year"] - divorce["dob_man"]
divorce["woman_age_marriage"] = divorce["marriage_year"] - divorce["dob_woma
```

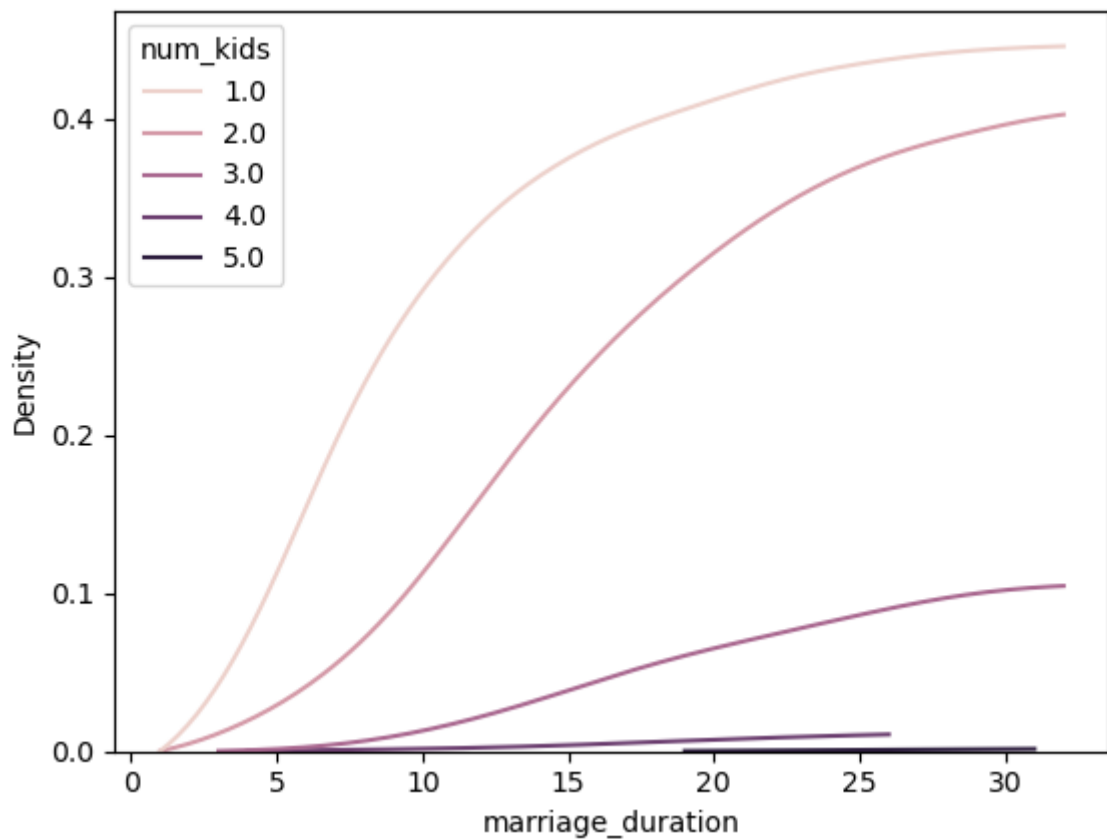
## 5

```
In [13]: sns.scatterplot(data = divorce , x = "woman_age_marriage" , y = "income_woma
plt.show()
```



## 6

```
In [64]: sns.kdeplot(data = divorce , x = "marriage_duration" , hue = "num_kids", cu  
plt.show()
```



# CLASS FREQUEnCY

```
In [65]: planes = pd.read_csv("Airlines_unclean.csv")
print(planes["Destination"].value_counts())
```

```
Cochin      4391
Banglore    2773
Delhi       1219
New Delhi   888
Hyderabad   673
Kolkata     369
Name: Destination, dtype: int64
```

```
In [66]: planes["Destination"].value_counts(normalize = True)
```

```
Out[66]: Cochin      0.425773
Banglore    0.268884
Delhi       0.118200
New Delhi   0.086105
Hyderabad   0.065257
Kolkata     0.035780
Name: Destination, dtype: float64
```

```
In [67]: pd.crosstab(planes["Source"] , planes["Destination"])
```

```
Out[67]: Destination  Bangalore  Cochin  Delhi  Hyderabad  Kolkata  New Delhi
Source
Banglore             0         0  1199             0         0         868
Chennai              0         0    0             0        364         0
Delhi                0      4318    0             0         0         0
Kolkata             2720         0    0             0         0         0
Mumbai              0         0    0             662         0         0
```

## Aggregated values with pd.crosstab()

```
In [68]: pd.crosstab(planes["Source"] , planes["Destination"] , values = planes["Price"])
```

```
Out[68]: Destination  Bangalore  Cochin  Delhi  Hyderabad  Kolkata  New Delhi
Source
Banglore             NaN      NaN  4823.0             NaN      NaN      10976.5
Chennai              NaN      NaN    NaN             NaN    3850.0         NaN
Delhi                NaN  10262.0    NaN             NaN      NaN         NaN
Kolkata             9345.0      NaN    NaN             NaN      NaN         NaN
Mumbai              NaN      NaN    NaN             3342.0      NaN         NaN
```

```
In [69]: salaries = pd.read_csv("Salary_Rupee_USD.csv", index_col = 0)
salaries["Job_Category"].value_counts(normalize = True)
```

Out[69]:

Data Science	0.277641
Data Engineering	0.272727
Data Analytics	0.226044
Machine Learning	0.120393
Other	0.068796
Managerial	0.034398

Name: Job\_Category, dtype: float64

## 8

```
In [70]: pd.crosstab(salaries["Company_Size"] , salaries["Experience"])
```

Out[70]:

Experience	EN	EX	MI	SE
Company_Size				
L	24	7	49	44
M	25	9	58	136
S	18	1	21	15

```
In [71]: pd.crosstab(salaries["Job_Category"] , salaries["Company_Size"])
```

Out[71]:

Company_Size		L	M	S
Job_Category				
Data Analytics		23	61	8
Data Engineering		28	72	11
Data Science		38	59	16
Machine Learning		17	19	13
Managerial		5	8	1
Other		13	9	6

```
In [72]: pd.crosstab(salaries["Job_Category"] , salaries["Company_Size"] , values = s
```

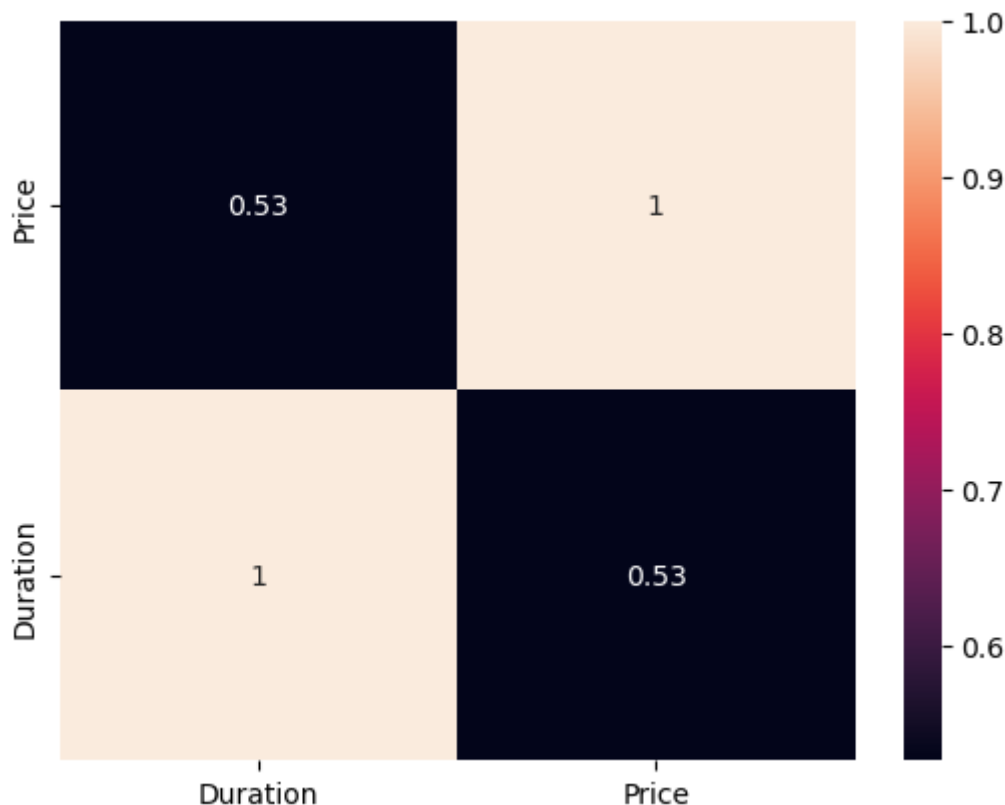
Out[72]:

Company_Size		L	M	S
Job_Category				
Data Analytics		112851.749217	95912.685246	53741.877000
Data Engineering		118939.035000	121287.060500	86927.136000
Data Science		96489.520105	116044.455864	62241.749250
Machine Learning		140779.491529	100794.236842	78812.586462
Managerial		190551.448800	150713.628000	31484.700000
Other		92873.911385	89750.578667	69871.248000

# CORRELATION

```
In [22]: planes = pd.read_csv('Airlines_unclean.csv', index_col = 0,
parse_dates=['Date_of_Journey', 'Dep_Time', 'Arrival_Time'])
# Remove the string character
planes["Duration"] = planes["Duration"].str.replace("h", ".")
planes["Duration"] = planes["Duration"].str.replace("m", "")
planes["Duration"] = planes["Duration"].str.replace(" ", "")
# Convert to float data type
planes["Duration"] = planes["Duration"].astype(float)
print(planes.info())
ax = sns.heatmap(planes.corr(), annot=True)
ax.set_ylim([0,2])
plt.show()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 10660 entries, 0 to 10659
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Airline                10233 non-null   object
1   Date_of_Journey        10338 non-null   datetime64[ns]
2   Source                 10473 non-null   object
3   Destination            10313 non-null   object
4   Route                  10404 non-null   object
5   Dep_Time               10400 non-null   datetime64[ns]
6   Arrival_Time           10466 non-null   datetime64[ns]
7   Duration                10446 non-null   float64
8   Total_Stops            10448 non-null   object
9   Additional_Info        10071 non-null   object
10  Price                  10044 non-null   float64
dtypes: datetime64[ns](3), float64(2), object(6)
memory usage: 999.4+ KB
None
```



## TOTAL STOPS

```
In [23]: print(planes["Total_Stops"].value_counts())
```

```
1 stop      5503
non-stop    3411
2 stops     1488
3 stops       45
4 stops        1
Name: Total_Stops, dtype: int64
```

```
In [25]: # planes["Total_Stops"] = planes["Total_Stops"].str.replace(" stops", "")
# planes["Total_Stops"] = planes["Total_Stops"].str.replace(" stop", "")
# planes["Total_Stops"] = planes["Total_Stops"].str.replace("non-stop", "0")
# planes["Total_Stops"] = planes["Total_Stops"].astype(int)
```

```
In [27]: # sns.heatmap(planes.corr(), annot = True)
# plt.show()
```

```
In [29]: # print(planes.dtypes)
```

## 9

```
In [58]: salaries = pd.read_csv("Salaries_with_date_of_response.csv", index_col = 0 ,
```



```
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '21/11/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '29/11/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '13/10/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '15/10/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '15/11/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '17/11/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '23/11/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '20/10/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '25/10/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '24/11/2020' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '20/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '13/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '24/11/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '25/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '30/11/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '17/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
```

```
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '20/11/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '17/11/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '26/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '28/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '29/11/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '23/10/2021' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '17/11/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '15/11/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '18/11/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '14/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '27/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '28/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '23/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '18/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '16/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '14/11/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
```

```

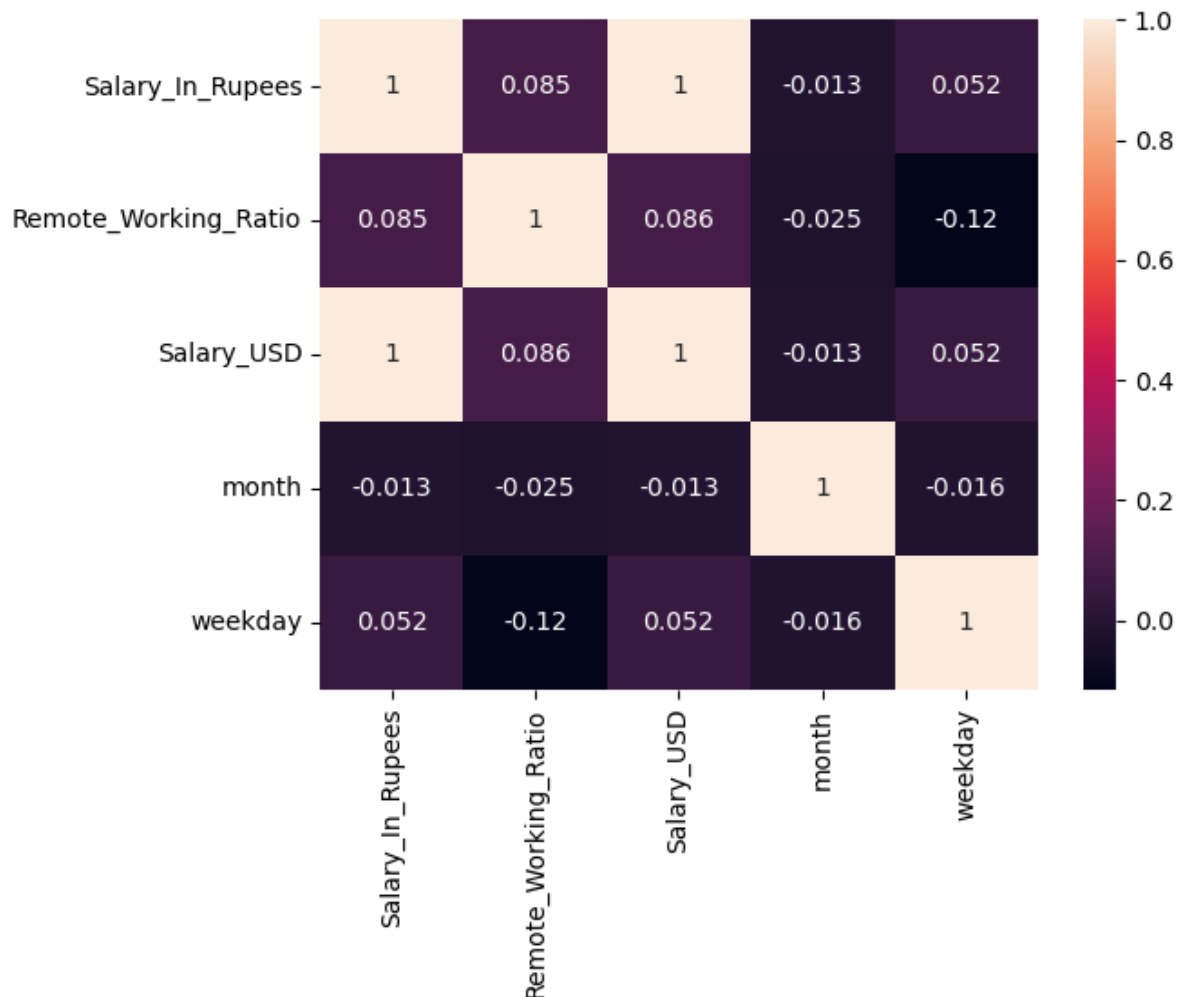
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '25/11/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '15/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '13/11/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(
/Users/richard/opt/anaconda3/lib/python3.9/site-packages/pandas/io/parsers/b
ase_parser.py:1070: UserWarning: Parsing '13/10/2022' in DD/MM/YYYY format.
Provide format or specify infer_datetime_format=True for consistent parsing.
    return tools.to_datetime(

```

```

In [59]: salaries["month"] = salaries["date_of_response"].dt.month
salaries["weekday"] = salaries["date_of_response"].dt.weekday
sns.heatmap(salaries.corr(), annot = True)
plt.show()

```



# 10

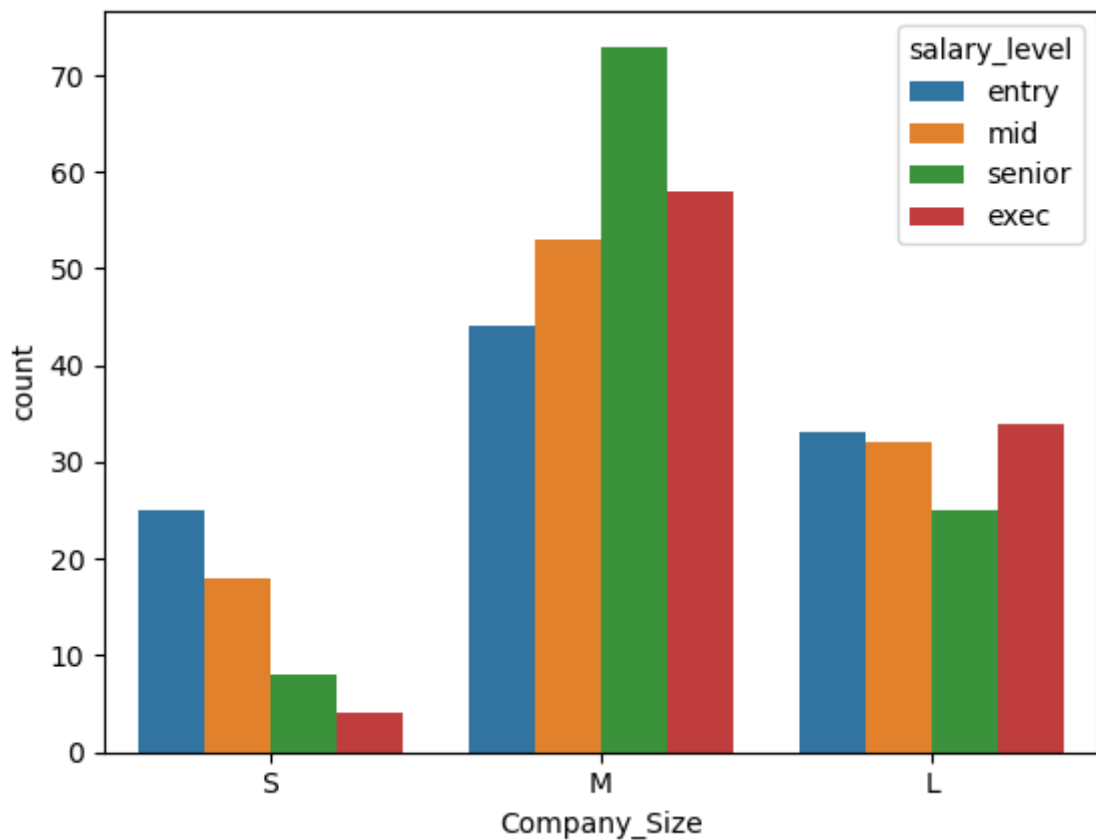
```

In [55]: salaries = pd.read_csv("Salaries_with_date_of_response.csv", index_col = 0)
twenty_fifth = salaries["Salary_USD"].quantile(0.25)
salaries_median = salaries["Salary_USD"].median()
seventy_fifth = salaries["Salary_USD"].quantile(0.75)
print(twenty_fifth, salaries_median,seventy_fifth)

```

60880.691999999995 97488.552 143225.1

```
In [56]: salary_labels = ["entry" , "mid" , "senior", "exec"]
salary_ranges = [0 ,twenty_fifth, salaries_median,seventy_fifth, salaries["S
salaries["salary_level"] = pd.cut(salaries["Salary_USD"], bins = salary_rang
sns.countplot(data=salaries , x = "Company_Size" , hue="salary_level")
plt.show()
```



## 11

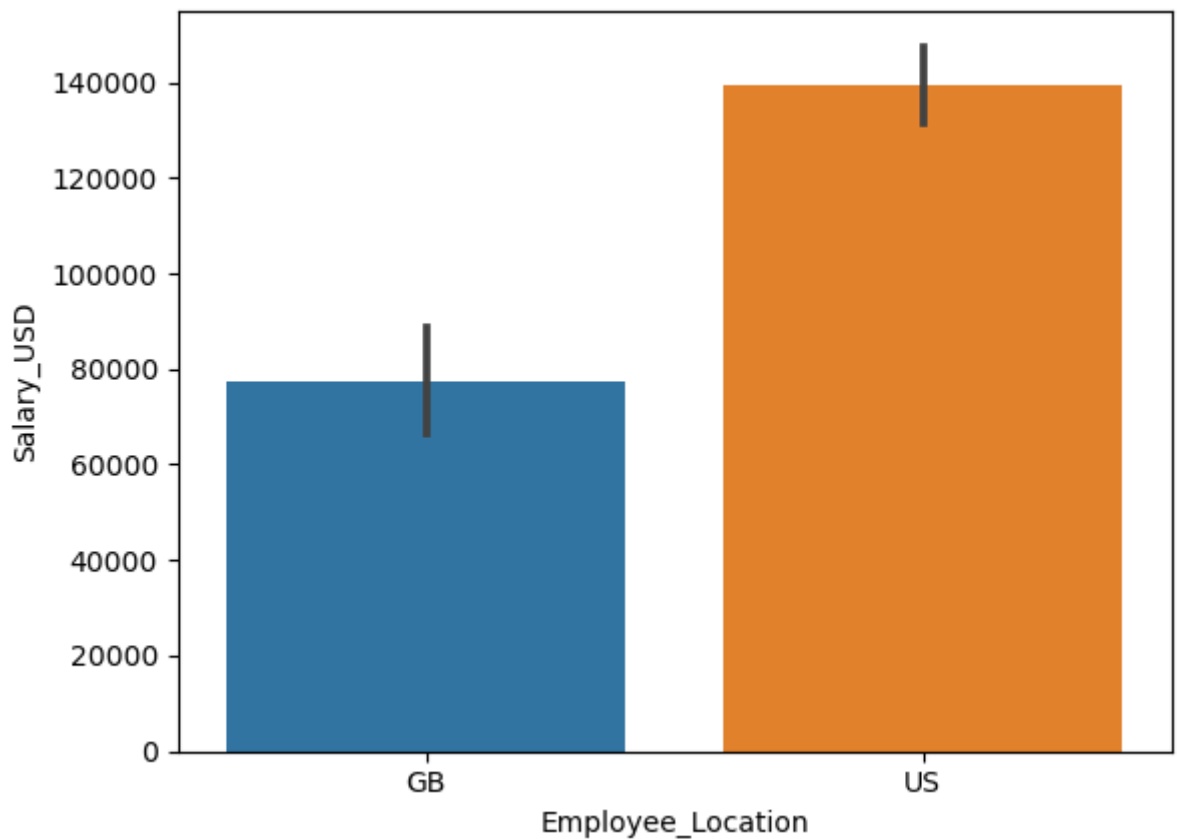
```
In [72]: usa = salaries[salaries["Employee_Location"].isin(["US", "GB"])]
usa
```

Out [72]:

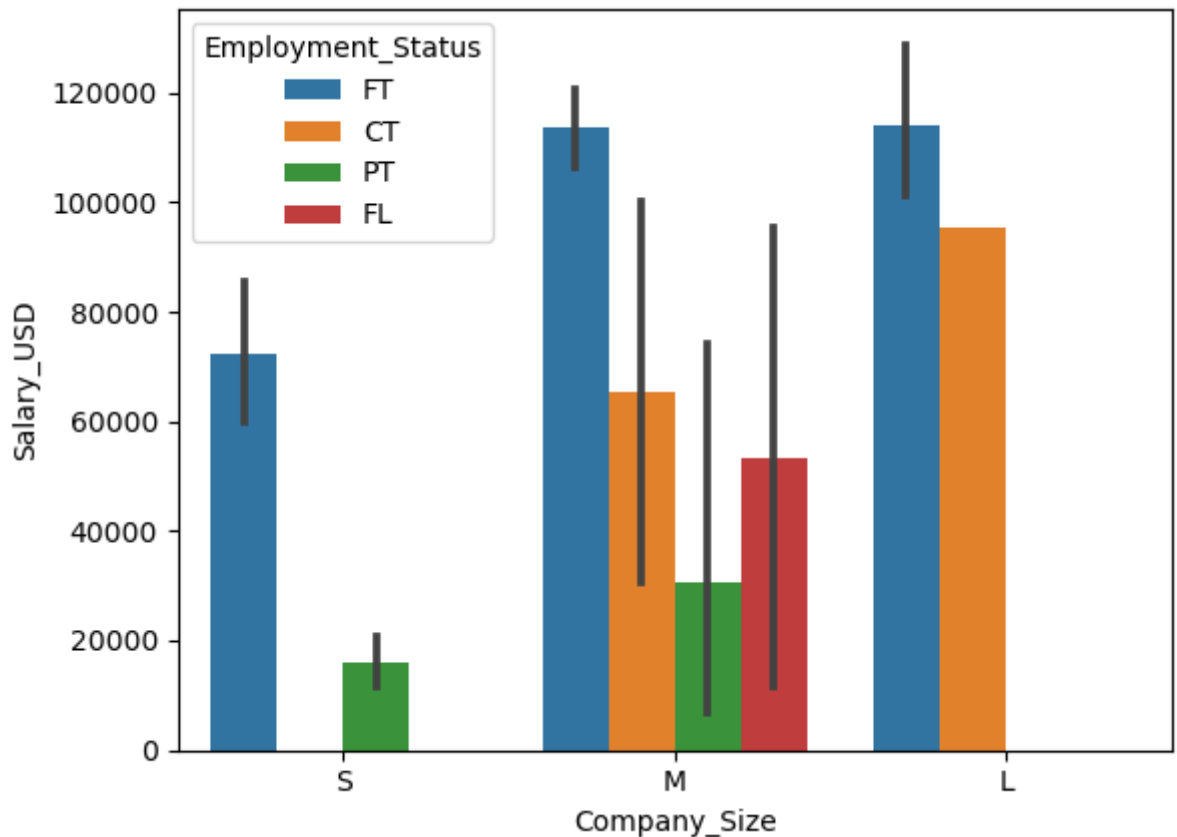
	Designation	date_of_response	Experience	Employment_Status	Salary_In_Rupees	Em
1	Big Data Engineer	2020-09-19	SE	FT	8680000.0	
3	Machine Learning Engineer	2020-11-29	SE	FT	11900000.0	
4	Data Analyst	2020-07-09	EN	FT	5730000.0	
5	Lead Data Scientist	2020-04-08	SE	FT	15100000.0	
7	Business Data Analyst	2020-10-13	MI	FT	10700000.0	
...	...	...	...	...	...	...
400	Data Scientist	2022-02-26	MI	FT	10300000.0	
402	Data Engineer	2022-10-27	SE	FT	12300000.0	
403	Data Engineer	2022-08-25	SE	FT	10000000.0	
404	Data Analyst	2022-11-08	SE	FT	10300000.0	
405	Data Analyst	2022-06-15	SE	FT	11900000.0	

257 rows x 13 columns

```
In [73]: sns.barplot(data = usa , x = "Employee_Location" , y = "Salary_USD")
plt.show()
```



```
In [75]: sns.barplot(data = salaries , y = "Salary_USD" , x= "Company_Size" , hue =  
Out[75]: <AxesSubplot:xlabel='Company_Size', ylabel='Salary_USD'>
```



The correct hypothesis is a. On average, small companies pay part-time employees less than large companies.

The plot shows that the salaries for part-time employees (PT) are lower for small companies (S) than for large companies (L). This suggests that there is a negative correlation between company size and salary for part-time employees. We can therefore hypothesize that, on average, small companies pay part-time employees less than large companies.

The other hypotheses are not supported by the data. For example, the plot does not show any difference in salary between freelancers (FL) at small and large companies. Additionally, the plot shows that the salaries for contractors (CT) are similar for medium-sized (M) and large companies.

Therefore, the correct hypothesis is a. On average, small companies pay part-time employees less than large companies. The answer is (a).

```
In [ ]:
```