ibm-attrition-3

June 23, 2025

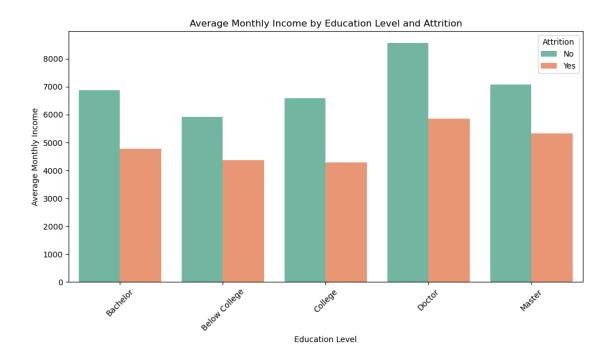
```
[20]: import pandas as pd
      df= pd.read_csv("ibm.csv")
 [8]:
     ibm
 [8]:
             Age Attrition
                                 BusinessTravel
                                                  DailyRate
                                                                           Department
      0
              41
                                  Travel_Rarely
                                                                                 Sales
                        Yes
                                                        1102
      1
              49
                             Travel_Frequently
                         No
                                                         279
                                                              Research & Development
      2
              37
                        Yes
                                  Travel_Rarely
                                                        1373
                                                              Research & Development
      3
              33
                             Travel_Frequently
                         No
                                                        1392
                                                              Research & Development
      4
              27
                         No
                                  Travel_Rarely
                                                         591
                                                              Research & Development
      1465
                             Travel_Frequently
                                                         884
                                                              Research & Development
              36
                         No
      1466
              39
                         No
                                  Travel_Rarely
                                                         613
                                                              Research & Development
      1467
                                  Travel_Rarely
                                                              Research & Development
              27
                         No
                                                         155
                             Travel_Frequently
      1468
              49
                         No
                                                        1023
                                                                                 Sales
              34
      1469
                                  Travel_Rarely
                                                         628
                                                              Research & Development
                         No
             DistanceFromHome
                                Education EducationField
                                                             EmployeeCount
      0
                             1
                                            Life Sciences
                                                                           1
                             8
                                             Life Sciences
      1
                                                                           1
      2
                             2
                                         2
                                                     Other
                                                                          1
      3
                             3
                                         4
                                             Life Sciences
                                                                           1
      4
                             2
                                         1
                                                   Medical
                                                                           1
                                         2
      1465
                            23
                                                   Medical
                                                                           1
      1466
                             6
                                         1
                                                   Medical
                                                                           1
                                         3
      1467
                             4
                                             Life Sciences
      1468
                             2
                                         3
                                                   Medical
                                                                           1
      1469
                             8
                                         3
                                                   Medical
                                                                           1
             EmployeeNumber
                                  RelationshipSatisfaction StandardHours
      0
                           1
      1
                           2
                                                           4
                                                                         80
                                                           2
      2
                           4
                                                                         80
      3
                           5
                                                           3
                                                                         80
      4
                           7
                                                           4
                                                                         80
```

 1465	2061		3	80
1466	2062		1	80
1467	2064		2	80
1468	2065		4	80
1469	2068 		1	80
	StockOptionLevel TotalW	lorkingYears	TrainingTimesL	astYear \
0	0	8	G	0
1	1	10		3
2	0	7		3
3	0	8		3
4	1	6		3
•••	•••	***	•••	
1465	1	17		3
1466	1	9		5
1467	1	6		0
1468	0	17		3
1469	0	6		3
	WorkLifeBalance YearsAtC	Company Years	InCurrentRole	\
0	1	6	4	•
1	3	10	7	
2				
	3	0	0	
3	3	8	7	
4	3	2	2	
•••			•••	
1465	3	5	2	
1466	3	7	7	
1467	3	6	2	
1468	2	9	6	
1469	4	4	3	
	YearsSinceLastPromotion	YearsWithCu	rrManager	
0	0	100121100	5	
1	1		7	
2	0		0	
3	3		0	
4	2		2	
•••	•••		***	
1465	0		3	
1466	1		7	
1467	0		3	
1468	0		8	
1469	1		2	
1100	1		-	

[1470 rows x 35 columns]

```
[38]: import matplotlib.pyplot as plt
      import seaborn as sns
      # Map education levels to their meanings
      education_mapping = {
          1: 'Below College',
          2: 'College',
          3: 'Bachelor',
         4: 'Master',
          5: 'Doctor'
      df['EducationLevel'] = df['Education'].map(education_mapping)
      \hbox{\it\# Group by EducationLevel and Attrition, then calculate average Monthly Income}
      avg_income = df.groupby(['EducationLevel', 'Attrition'])['MonthlyIncome'].
       →mean().reset_index()
      # Optional: Format MonthlyIncome for better readability
      avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
      print(avg_income)
      plt.figure(figsize=(10, 6))
      sns.barplot(data=avg_income, x='EducationLevel', y='MonthlyIncome', u
       ⇔hue='Attrition', palette='Set2')
      # Add labels and title
      plt.title('Average Monthly Income by Education Level and Attrition')
      plt.xlabel('Education Level')
      plt.ylabel('Average Monthly Income')
      plt.xticks(rotation=45)
      plt.tight_layout()
      # Show the plot
      plt.show()
```

	Educati	ionLevel	Attrition	MonthlyIncome
0	I	Bachelor	No	6882.92
1	I	Bachelor	Yes	4770.24
2	Below	College	No	5926.13
3	Below	College	Yes	4360.16
4		College	No	6586.06
5		College	Yes	4282.55
6		Doctor	No	8559.91
7		Doctor	Yes	5850.20
8		Master	No	7087.81
9		Master	Yes	5335.16



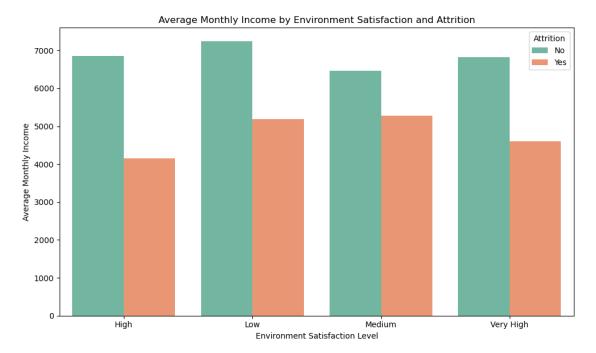
```
[40]: import pandas as pd
      # Map education levels to their meanings
      environmental_mapping = {
      1: 'Low',
      2:'Medium',
      3: 'High',
      4:'Very High'
      df['SatisfactionLevel'] = df['EnvironmentSatisfaction'].
       →map(enviromental_mapping)
      # Group by EducationLevel and Attrition, then calculate average MonthlyIncome
      avg_income = df.groupby(['SatisfactionLevel', 'Attrition'])['MonthlyIncome'].
       →mean().reset_index()
      # Optional: Format MonthlyIncome for better readability
      avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
      print(avg_income)
      plt.figure(figsize=(10, 6))
```

```
sns.barplot(data=avg_income, x='SatisfactionLevel', y='MonthlyIncome',
hue='Attrition', palette='Set2')

# Add title and labels
plt.title('Average Monthly Income by Environment Satisfaction and Attrition')
plt.xlabel('Environment Satisfaction Level')
plt.ylabel('Average Monthly Income')
plt.ylabel('Average Monthly Income')
plt.xticks(rotation=0)
plt.tight_layout()

# Show plot
plt.show()
```

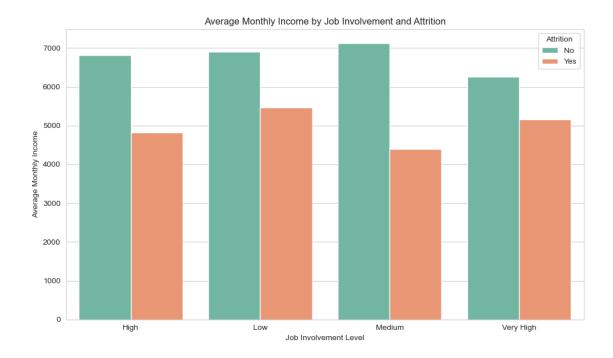
	${\tt SatisfactionLevel}$	Attrition	MonthlyIncome
0	High	No	6851.10
1	High	Yes	4156.94
2	Low	No	7236.89
3	Low	Yes	5186.53
4	Medium	No	6460.00
5	Medium	Yes	5283.51
6	Very High	No	6827.79
7	Very High	Yes	4603.17



```
[42]: # Map education levels to their meanings
JobInvolvement_mapping = {
```

```
1: 'Low',
2:'Medium',
3:'High',
4:'Very High'
df['jobInvolve'] = df['JobInvolvement'].map(JobInvolvement_mapping)
# Group by EducationLevel and Attrition, then calculate average MonthlyIncome
avg_income = df.groupby(['jobInvolve', 'Attrition'])['MonthlyIncome'].mean().
⇔reset_index()
# Optional: Format MonthlyIncome for better readability
avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
print(avg_income)
sns.set_style("whitegrid")
# Create the grouped bar chart
plt.figure(figsize=(10, 6))
sns.barplot(data=avg_income, x='jobInvolve', y='MonthlyIncome',
 ⇔hue='Attrition', palette='Set2')
# Add labels and title
plt.title('Average Monthly Income by Job Involvement and Attrition')
plt.xlabel('Job Involvement Level')
plt.ylabel('Average Monthly Income')
plt.xticks(rotation=0)
plt.tight_layout()
# Show the plot
plt.show()
```

	${\tt jobInvolve}$	Attrition	MonthlyIncome
0	High	No	6808.26
1	High	Yes	4817.93
2	Low	No	6902.98
3	Low	Yes	5465.68
4	Medium	No	7126.31
5	Medium	Yes	4397.46
6	Very High	No	6260.81
7	Very High	Yes	5157.00



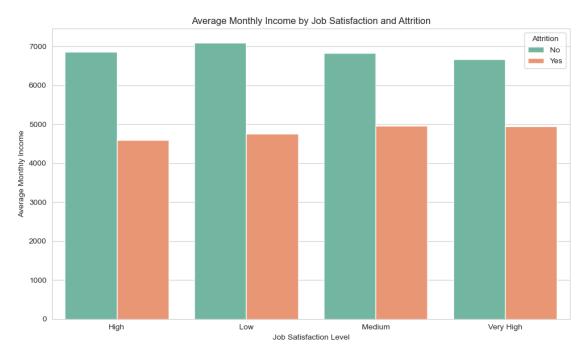
```
[44]: # Map education levels to their meanings
      JobSatisfaction_mapping = {
      1: 'Low',
      2:'Medium',
      3: 'High',
      4:'Very High'
      df['jobSatify'] = df['JobSatisfaction'].map(JobSatisfaction_mapping)
      # Group by EducationLevel and Attrition, then calculate average MonthlyIncome
      avg_income = df.groupby(['jobSatify', 'Attrition'])['MonthlyIncome'].mean().
       →reset_index()
      # Optional: Format MonthlyIncome for better readability
      avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
      print(avg_income)
      # Set style
      sns.set_style("whitegrid")
      # Create clustered bar plot
      plt.figure(figsize=(10, 6))
```

```
sns.barplot(data=avg_income, x='jobSatify', y='MonthlyIncome', hue='Attrition', upalette='Set2')

# Add labels and title
plt.title('Average Monthly Income by Job Satisfaction and Attrition')
plt.xlabel('Job Satisfaction Level')
plt.ylabel('Average Monthly Income')
plt.ylabel('Average Monthly Income')
plt.xticks(rotation=0)
plt.tight_layout()

# Show the plot
plt.show()
```

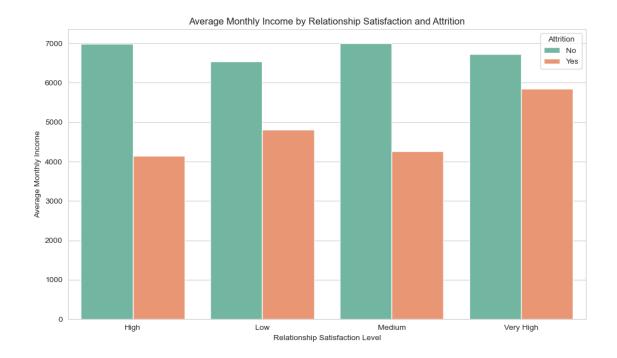
	jobSatify	Attrition	MonthlyIncome
0	High	No	6853.31
1	High	Yes	4595.99
2	Low	No	7096.90
3	Low	Yes	4752.80
4	Medium	No	6834.89
5	Medium	Yes	4962.76
6	Very High	No	6668.11
7	Very High	Yes	4943.50



```
[46]: # Map education levels to their meaning
RelationshipSatisfaction_mapping = {
```

```
1: 'Low',
2:'Medium',
3:'High',
4:'Very High'
}
df['RelationSatify'] = df['RelationshipSatisfaction'].
 map(RelationshipSatisfaction_mapping)
# Group by EducationLevel and Attrition, then calculate average MonthlyIncome
avg_income = df.groupby(['RelationSatify', 'Attrition'])['MonthlyIncome'].
→mean().reset_index()
# Optional: Format MonthlyIncome for better readability
avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
print(avg_income)
sns.set_style("whitegrid")
# Create clustered bar plot
plt.figure(figsize=(10, 6))
sns.barplot(data=avg_income, x='RelationSatify', y='MonthlyIncome', u
 ⇔hue='Attrition', palette='Set2')
# Add labels and title
plt.title('Average Monthly Income by Relationship Satisfaction and Attrition')
plt.xlabel('Relationship Satisfaction Level')
plt.ylabel('Average Monthly Income')
plt.xticks(rotation=0)
plt.tight_layout()
# Show the plot
plt.show()
```

	RelationSatify	Attrition	MonthlyIncome
0	High	No	6981.96
1	High	Yes	4149.17
2	Low	No	6534.10
3	Low	Yes	4808.82
4	Medium	No	7005.07
5	Medium	Yes	4257.27
6	Very High	No	6732.31
7	Very High	Yes	5847.97



```
[48]: # Map education levels to their meanin
      PerformanceRating_mapping = {
      1 : 'Low',
      2: 'Good',
      3 : 'Excellent',
      4 : 'Outstanding',
      df['Rating'] = df['JobSatisfaction'].map(PerformanceRating_mapping)
      # Group by EducationLevel and Attrition, then calculate average MonthlyIncome
      avg_income = df.groupby(['Rating', 'Attrition'])['MonthlyIncome'].mean().
       →reset_index()
      # Optional: Format MonthlyIncome for better readability
      avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
      print(avg_income)
      sns.set_style("whitegrid")
      # Clustered bar plot
      plt.figure(figsize=(10, 6))
      sns.barplot(data=avg_income, x='Rating', y='MonthlyIncome', hue='Attrition', u
       →palette='Set2')
```

```
# Add title and labels
plt.title('Average Monthly Income by Performance Rating and Attrition')
plt.xlabel('Performance Rating')
plt.ylabel('Average Monthly Income')
plt.xticks(rotation=0)
plt.tight_layout()

# Show plot
plt.show()
```

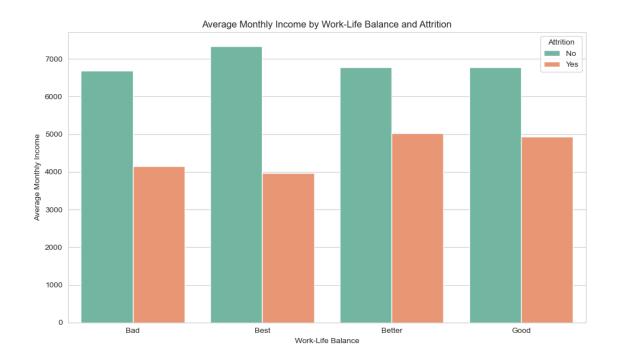
	Rating	Attrition	MonthlyIncome
0	Excellent	No	6853.31
1	Excellent	Yes	4595.99
2	Good	No	6834.89
3	Good	Yes	4962.76
4	Low	No	7096.90
5	Low	Yes	4752.80
6	Outstanding	No	6668.11
7	Outstanding	Yes	4943.50



```
[50]: # Map education levels to their meanin
WorkLifeBalance_mapping = {
    1: 'Bad',
    2: 'Good',
    3: 'Better',
```

```
4: 'Best',
}
df['WorkBalance'] = df['WorkLifeBalance'].map(WorkLifeBalance_mapping)
# Group by EducationLevel and Attrition, then calculate average MonthlyIncome
avg_income = df.groupby(['WorkBalance', 'Attrition'])['MonthlyIncome'].mean().
 →reset_index()
# Optional: Format MonthlyIncome for better readability
avg_income['MonthlyIncome'] = avg_income['MonthlyIncome'].round(2)
print(avg_income)
sns.set_style("whitegrid")
# Plot
plt.figure(figsize=(10, 6))
sns.barplot(data=avg_income, x='WorkBalance', y='MonthlyIncome', u
 ⇔hue='Attrition', palette='Set2')
plt.title('Average Monthly Income by Work-Life Balance and Attrition')
plt.xlabel('Work-Life Balance')
plt.ylabel('Average Monthly Income')
plt.xticks(rotation=0)
plt.tight_layout()
plt.show()
```

	${\tt WorkBalance}$	Attrition	MonthlyIncome
0	Bad	No	6679.67
1	Bad	Yes	4143.56
2	Best	No	7340.17
3	Best	Yes	3975.19
4	Better	No	6782.62
5	Better	Yes	5022.01
6	Good	No	6772.85
7	Good	Yes	4928.05



[]: