

# HW 2: Types of Data

CPE232 Data Models

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## Import Dependency

```
In [100... import pandas as pd
import matplotlib.pyplot as plt
```

## Part 1 : Basic Python

### Task 1

เขียนโปรแกรมสำหรับใส่ค่า Input แต่ละ type ที่ต่างกัน ทำการแสดง type ของแต่ละตัว และทดลองแปลงค่า Float ให้กลายเป็น int พร้อมแสดง Before After

```
In [102... # Input values from the user
string_value = input("Enter a string: ")
integer_value = int(input("Enter an integer: "))
float_value = float(input("Enter a float: "))
```

```
In [108... # Display the data types
print(f"string_value type : {type(string_value)}")
print(f"integer_value type : {type(integer_value)}")
print(f"float_value type : {type(float_value)}")
```

```
string_value type : <class 'str'>
integer_value type : <class 'int'>
float_value type : <class 'float'>
```

```
In [109... # Convert integer to float and vice versa
print(f"Before converting : {integer_value} , {type(integer_value)}")
integer_To_float_value = float(integer_value)
print(f"After converting : {integer_To_float_value}, {type(integer_To_float_value)}")
```

```
Before converting : 37 , <class 'int'>
After converting : 37.0, <class 'float'>
```

```
In [110... # Show before and after convert float to int
print(f"Before converting : {float_value} , {type(float_value)}")
float_To_integer_value = int(float_value)
print(f"After converting : {float_To_integer_value}, {type(float_To_integer_value)}")
```

```
Before converting : 3.14 , <class 'float'>
After converting : 3, <class 'int'>
```

### Task 2

มีคะแนนของนักเรียนแต่ละคนให้ในรูปแบบ Dictionary จงหาคะแนนเฉลี่ยของนักเรียนแต่ละคนในทุกวิชาและหาคนที่ได้คะแนนเฉลี่ยสูงสุด

In [111...

```
students_grades = {  
    "John": [85, 90, 78],  
    "Alice": [88, 92, 80],  
    "Bob": [75, 85, 72],  
    "Diana": [90, 95, 94],  
    "Charlie": [70, 65, 80]  
}
```

In [112...

```
# Calculate and print the average grade for each student  
for name, grades in students_grades.items():  
    avg_grade = sum(grades) / len(grades)  
    print(f"{name} average grade : {avg_grade}")
```

John average grade : 84.33333333333333  
Alice average grade : 86.66666666666667  
Bob average grade : 77.33333333333333  
Diana average grade : 93.0  
Charlie average grade : 71.66666666666667

In [113...

```
# Find the student with the highest average grade  
avg_grade = {name : sum(grades)/len(grades) for name, grades in students_grades.items()}  
  
top_student_name = max(avg_grade, key=avg_grade.get)  
top_student_avg_grade = avg_grade[top_student_name]  
  
print(f"The student with the highest average grade is {top_student_name} with an average of
```

The student with the highest average grade is Diana with an average of 93.0

## Part 2 : Working with CSV!

จงสร้างคอลัมน์ใหม่ให้กับ CSV ไฟล์นี้ชื่อว่า Bonus และให้คำนวณ Bonus ให้กับพนักงานที่อยู่แผนก Sales เป็นจำนวน 10% จาก MonthlyRate ของคนๆนั้น

In [117...

```
# Load the CSV file  
file_path = './sources/employee_data.csv'  
df = pd.read_csv(file_path)
```

In [118...

```
df.head()
```

Out[118...

	Age	DailyRate	Department	EducationField	Gender	MaritalStatus	MonthlyRate	OverTime
0	41	1102	Sales	Life Sciences	Female	Single	19479	Yes
1	49	279	Research & Development	Life Sciences	Male	Married	24907	No
2	37	1373	Research & Development	Other	Male	Single	2396	Yes
3	33	1392	Research & Development	Life Sciences	Female	Married	23159	Yes
4	27	591	Research & Development	Medical	Male	Married	16632	No

In [119...

```
# Add new column call 'Bonus'  
df['Bonus'] = 0.0
```

In [120...

```
# Calculate 10% of bonus into the 'Bonus' Column .
df["Bonus"] = df["MonthlyRate"] * 0.1 * (df["Department"] == "Sales")
```

In [121...

```
# Print and show result
df
```

Out[121...

	Age	DailyRate	Department	EducationField	Gender	MaritalStatus	MonthlyRate	OverTime
0	41	1102	Sales	Life Sciences	Female	Single	19479	Yes
1	49	279	Research & Development	Life Sciences	Male	Married	24907	No
2	37	1373	Research & Development	Other	Male	Single	2396	Yes
3	33	1392	Research & Development	Life Sciences	Female	Married	23159	Yes
4	27	591	Research & Development	Medical	Male	Married	16632	No
...	...	...	...	...	...	...	...	...
1465	36	884	Research & Development	Medical	Male	Married	12290	No
1466	39	613	Research & Development	Medical	Male	Married	21457	No
1467	27	155	Research & Development	Life Sciences	Male	Married	5174	Yes
1468	49	1023	Sales	Medical	Male	Married	13243	No
1469	34	628	Research & Development	Medical	Male	Married	10228	No

1470 rows × 9 columns



In [122...

```
# Save the updated DataFrame back to a CSV file
df.to_csv('./sources/employee_data.csv', index=False)
```

## Part 3 : Working with Matplotlib

จงพล็อตกราฟแสดงค่าเฉลี่ยเงินเดือนของพนักงานในแต่ละ Department เพื่อเปรียบเทียบเงินเดือนเฉลี่ยของแต่ละ Department

In [123...

```
# Load CSV and create DataFrame
data = pd.read_csv('./sources/employee_data.csv')
df = pd.DataFrame(data)
```

In [124...

```
# Calculate average MonthlyRate by Department
avg_monthly_rate_by_department = df.groupby('Department')['MonthlyRate'].mean()

avg_monthly_rate_by_department
```

```
Out[124... Department
Human Resources      13492.984127
Research & Development 14284.865765
Sales                14489.793722
Name: MonthlyRate, dtype: float64
```

```
In [125... df['Department'].unique()
```

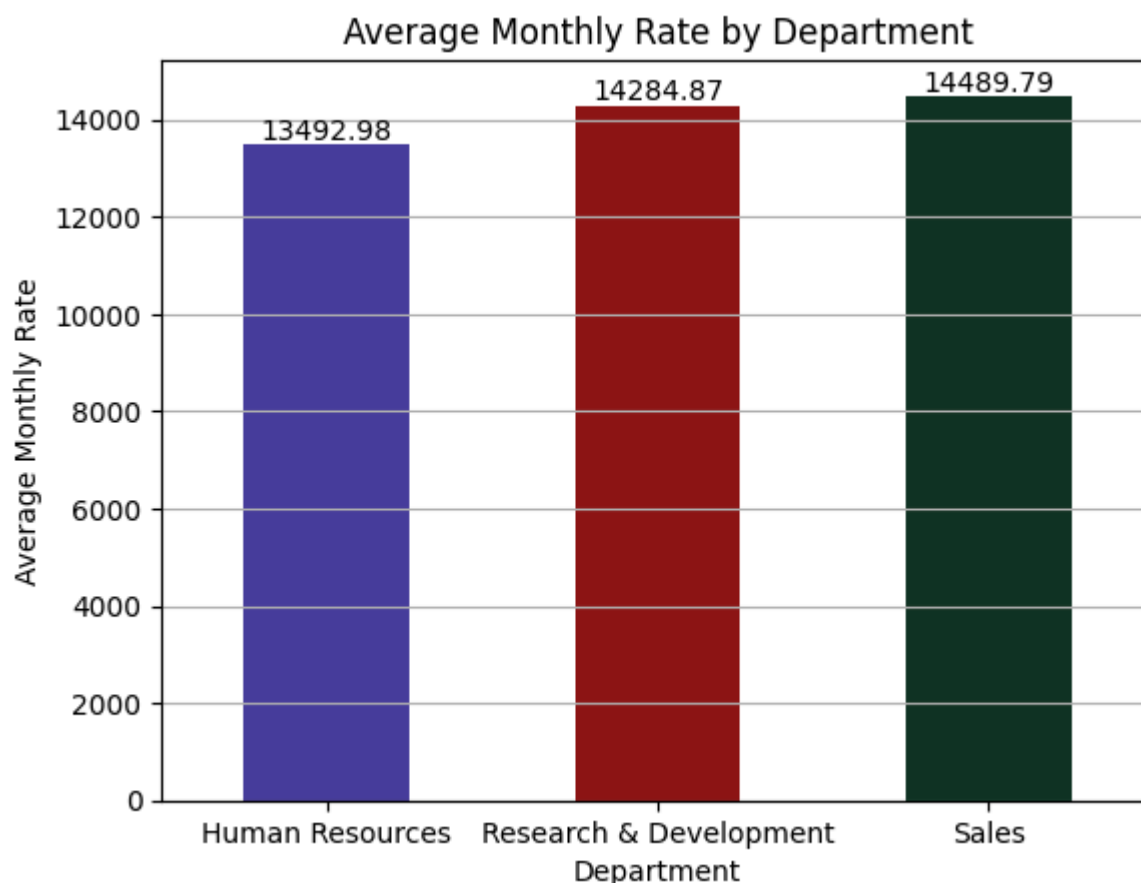
```
Out[125... array(['Sales', 'Research & Development', 'Human Resources'], dtype=object)
```

```
In [127... # Plotting the bar chart
avg_monthly_rate_by_department.plot(kind='bar', color=['#493D9E', '#8E1616', '#123524'])

plt.title('Average Monthly Rate by Department')
plt.xlabel('Department')
plt.ylabel('Average Monthly Rate')
plt.xticks(rotation=0)
plt.grid(axis='y')

for index, value in enumerate(avg_monthly_rate_by_department):
    plt.text(index, value+100, str(round(float(value),2)), ha='center')

plt.show()
```



## Challenge!!! : Working with SATAN (Optional)

ดีมากเหล่าเด็กๆที่ยังไม่ยอมแพ้ให้กับวิชานี้!



เรื่องมันมีอยู่ว่า จากไฟล์ CSV ที่พี่ไทให้น้องๆไป...มันไม่ใช่ไฟล์ธรรมดา แต่มันเป็นไฟล์ข้อมูลในองค์กรหนึ่งที่ Alya-San นั้นได้ทำงานอยู่



ซึ่งคุณอาเรารู้สึกว่ามีสิ่งไม่ชอบมาพากลภายในบริษัท ว่ารายได้ของพนักงานชายและหญิงมีความไม่เท่าเทียมกัน

รวมถึงรัฐอาจมีการแทรกแซงเพื่อให้ประชากรภายในประเทศเพิ่มสูงขึ้น รัฐจึงแอบเพิ่มเงินเดือนให้กับผู้ที่แต่งงานอยู่ริเปล่าเมื่อเทียบกับคนโสด

และคนที่จบจากสาขาที่แตกต่างกัน พวกเขาได้เงินเดือนมากกว่ากันมากน้อยแค่ไหน

น้องๆช่วยคุณอาเรย์พิสูจน์หน่อยว่าคุณอาเรย์ไปเองหรือมันเป็นเรื่องจริง!

เพราะสำหรับคุณอาเรย์แล้ว ทุกคนล้วน เ ท่ า เ ที ย ม กั น

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## Data Analysis & Visualization

In [128...

```
# Load CSV and create DataFrame
df = pd.read_csv('./sources/employee_data.csv')
df
```

Out[128...

	Age	DailyRate	Department	EducationField	Gender	MaritalStatus	MonthlyRate	OverTime
0	41	1102	Sales	Life Sciences	Female	Single	19479	Yes
1	49	279	Research & Development	Life Sciences	Male	Married	24907	No
2	37	1373	Research & Development	Other	Male	Single	2396	Yes
3	33	1392	Research & Development	Life Sciences	Female	Married	23159	Yes
4	27	591	Research & Development	Medical	Male	Married	16632	No
...	...	...	...	...	...	...	...	...
1465	36	884	Research & Development	Medical	Male	Married	12290	No
1466	39	613	Research & Development	Medical	Male	Married	21457	No
1467	27	155	Research & Development	Life Sciences	Male	Married	5174	Yes
1468	49	1023	Sales	Medical	Male	Married	13243	No
1469	34	628	Research & Development	Medical	Male	Married	10228	No

1470 rows × 9 columns



In [129...

```
# Create Total Income (MonthlyRate + Bonus)
df['TotalIncome'] = df['MonthlyRate'] + df['Bonus']
df
```

Out[129...

	Age	DailyRate	Department	EducationField	Gender	MaritalStatus	MonthlyRate	OverTime
0	41	1102	Sales	Life Sciences	Female	Single	19479	Yes
1	49	279	Research & Development	Life Sciences	Male	Married	24907	No
2	37	1373	Research & Development	Other	Male	Single	2396	Yes
3	33	1392	Research & Development	Life Sciences	Female	Married	23159	Yes
4	27	591	Research & Development	Medical	Male	Married	16632	No
...	...	...	...	...	...	...	...	...
1465	36	884	Research & Development	Medical	Male	Married	12290	No
1466	39	613	Research & Development	Medical	Male	Married	21457	No
1467	27	155	Research & Development	Life Sciences	Male	Married	5174	Yes
1468	49	1023	Sales	Medical	Male	Married	13243	No
1469	34	628	Research & Development	Medical	Male	Married	10228	No

1470 rows × 10 columns



## Determine if the total income for males and females are unequal ?

In [130...

```
df['Gender'].unique()
```

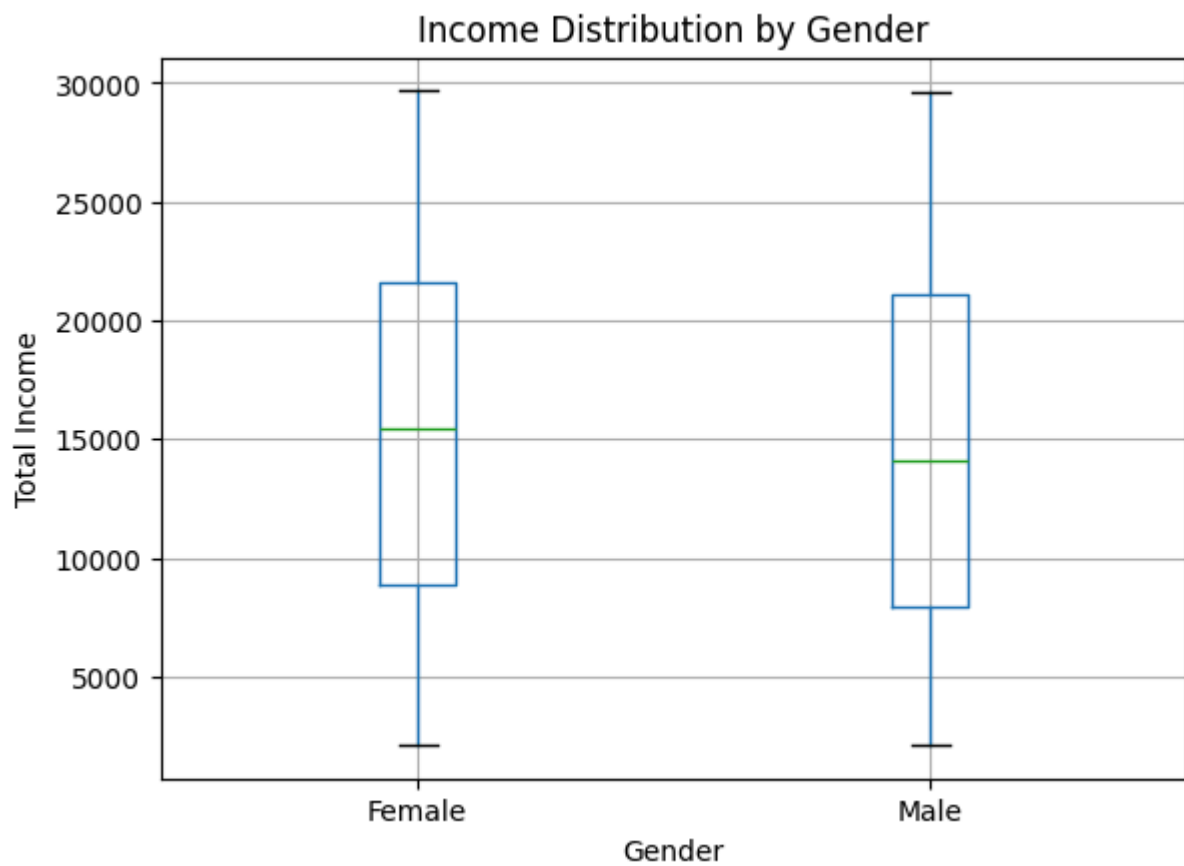
Out[130...

```
array(['Female', 'Male'], dtype=object)
```

In [131...

```
# Boxplot for salary distribution by gender
plt.figure(figsize=(6, 5))
df.boxplot(column='TotalIncome', by='Gender', grid=True)
plt.title("Income Distribution by Gender")
plt.xlabel("Gender")
plt.ylabel("Total Income")
plt.suptitle("")
plt.show()
```

<Figure size 600x500 with 0 Axes>

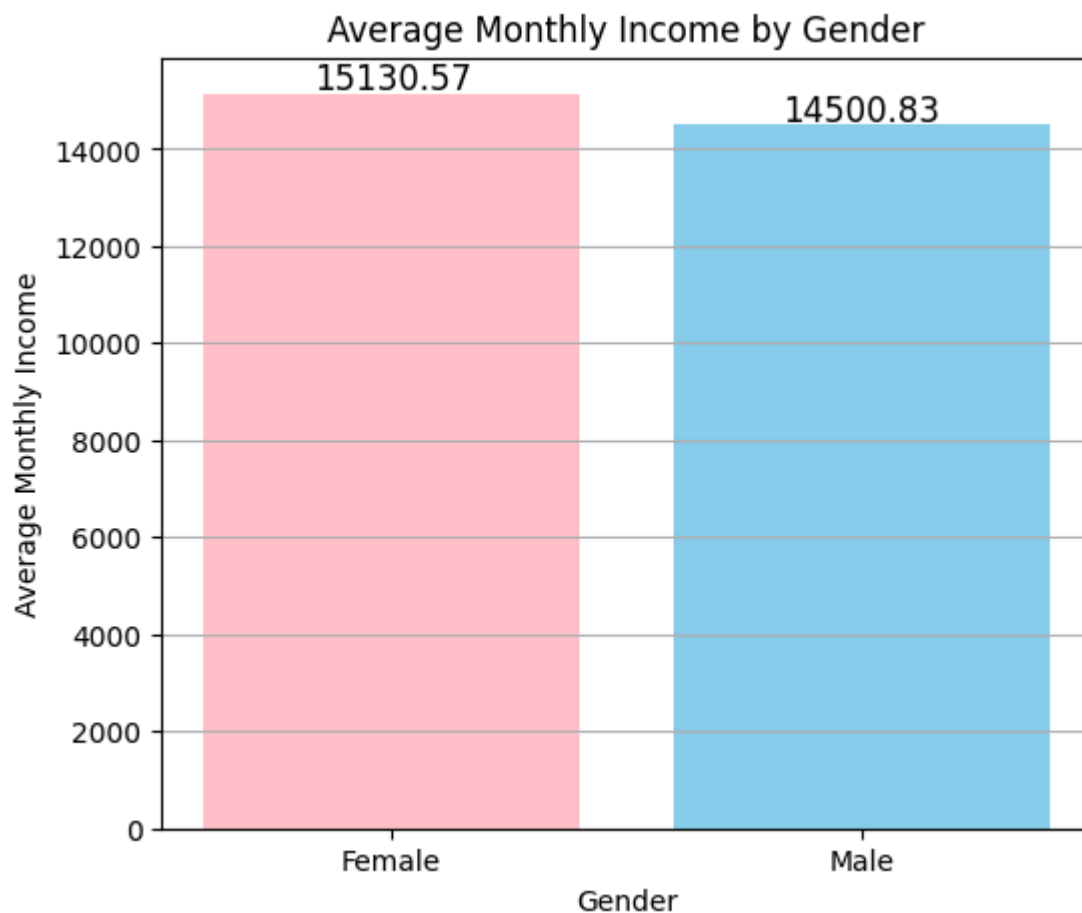


```
In [132...] avg_income_by_gender = df.groupby('Gender')['TotalIncome'].mean()  
avg_income_by_gender
```

```
Out[132...] Gender  
Female      15130.568707  
Male        14500.830385  
Name: TotalIncome, dtype: float64
```

```
In [133...] # Plot bar chart  
plt.figure(figsize=(6, 5))  
plt.bar(avg_income_by_gender.index, avg_income_by_gender.values, color=['pink', 'skyblue'])  
plt.title("Average Monthly Income by Gender")  
plt.xlabel("Gender")  
plt.ylabel("Average Monthly Income")  
plt.grid(axis='y')  
  
# Add values on bars  
for i, v in enumerate(avg_income_by_gender.values):  
    plt.text(i, v + 100, str(round(v, 2)), ha='center', fontsize=12)  
  
plt.show()
```





**Determine if the monthly rate for males and females are unequal ?**

**DIAGNOS :** Since the box plot shows that the total income distribution of both **male** and **female** employees is similarly grouped, I decided to compare the data using the **average total income**. The bar chart reveals that, on average, **female employees** earn **\$15,130.57**, while **male employees** earn **\$14,500.83**, making the **female bar** slightly higher than the **male bar**. This suggests that, overall, **female employees** tend to have a higher total income than their **male counterparts**. ✓

**Determine if the state secretly increases the monthly rates of married individuals compared to those who are single ?**

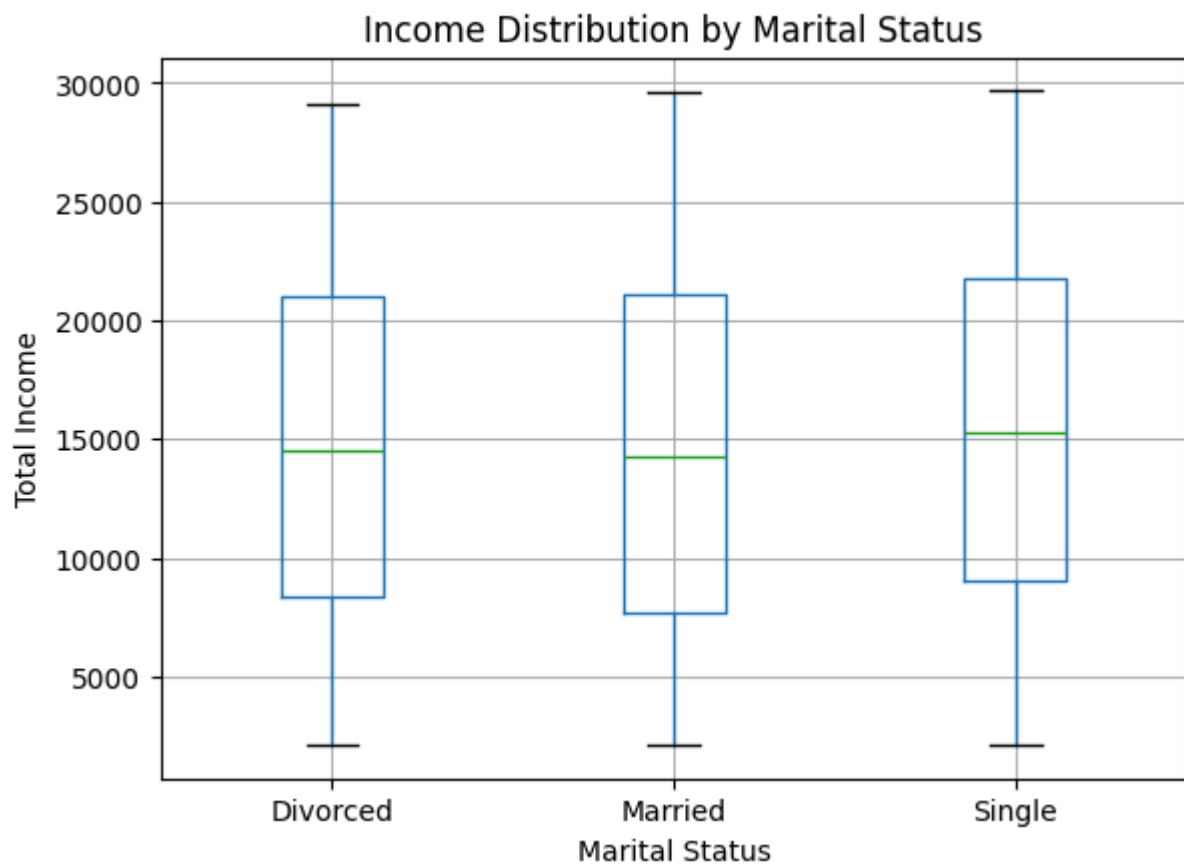
In [134... `df['MaritalStatus'].unique()`

Out[134... `array(['Single', 'Married', 'Divorced'], dtype=object)`

In [135... 

```
# Boxplot for salary distribution by Marital Status
plt.figure(figsize=(6, 5))
df.boxplot(column='TotalIncome', by='MaritalStatus', grid=True)
plt.title("Income Distribution by Marital Status")
plt.xlabel("Marital Status")
plt.ylabel("Total Income")
plt.suptitle("")
plt.show()
```

<Figure size 600x500 with 0 Axes>



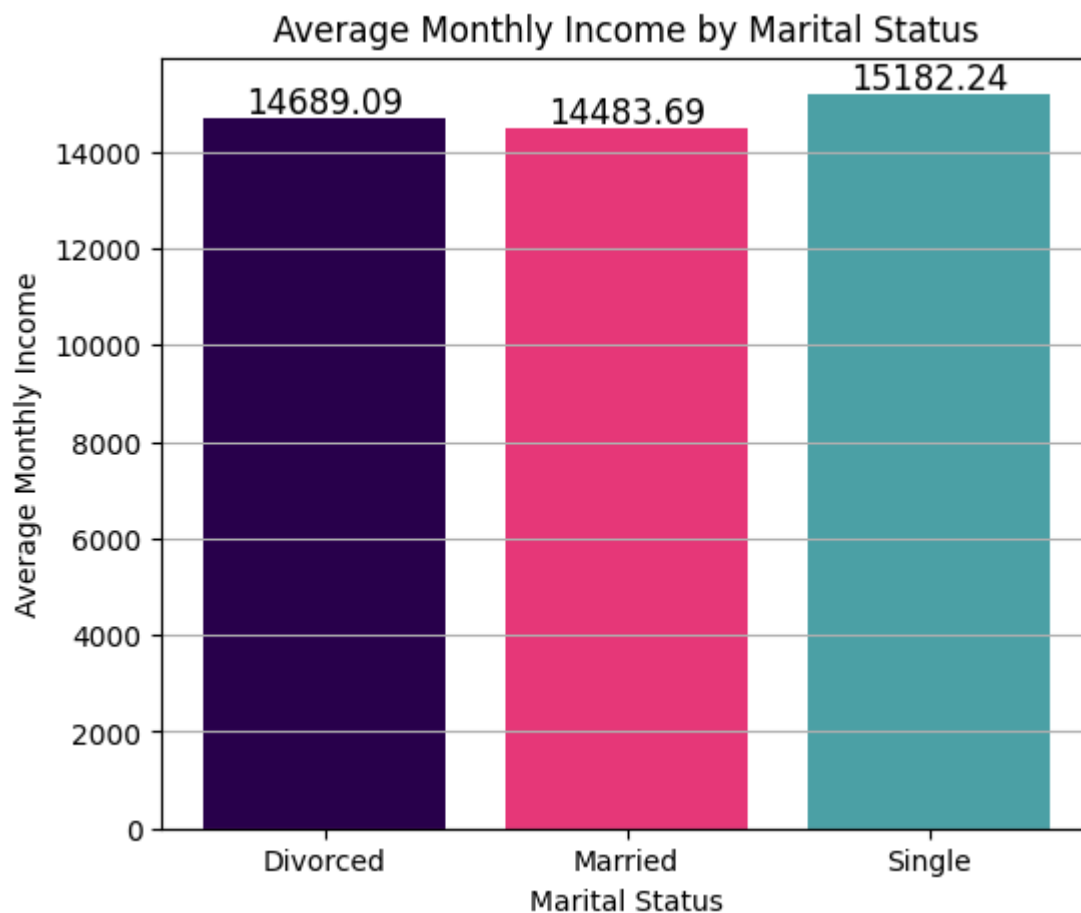
```
In [136...] avg_income_by_status = df.groupby('MaritalStatus')['TotalIncome'].mean()
avg_income_by_status
```

```
Out[136...] MaritalStatus
Divorced    14689.092355
Married     14483.687964
Single      15182.237447
Name: TotalIncome, dtype: float64
```

```
In [137...] # Plot bar chart
plt.figure(figsize=(6, 5))
plt.bar(avg_income_by_status.index, avg_income_by_status.values, color=['#2A004E', '#E73879'])
plt.title("Average Monthly Income by Marital Status")
plt.xlabel("Marital Status")
plt.ylabel("Average Monthly Income")
plt.grid(axis='y')

# Add values on bars
for i, v in enumerate(avg_income_by_status.values):
    plt.text(i, v + 100, str(round(v, 2)), ha='center', fontsize=12)

plt.show()
```



**Determine if the state secretly increases the monthly rates of married individuals compared to those who are single ?**

**DIAGNOS :** Since the box plot shows that the total income distribution of **single** and **married** employees is similarly grouped, and if divorced individuals are excluded from the single category, I decided to compare the data using the **average total income**. The bar chart reveals that, on average, **single employees** earn **\$15,182.24**, divorced employees earn **\$14,689.09**, and **married employees** earn **\$14,483.69**, making the **single group** slightly higher than the others. This suggests that there is no clear evidence that the state is secretly increasing the monthly wages of **married individuals** compared to those who are **single**. 🤔

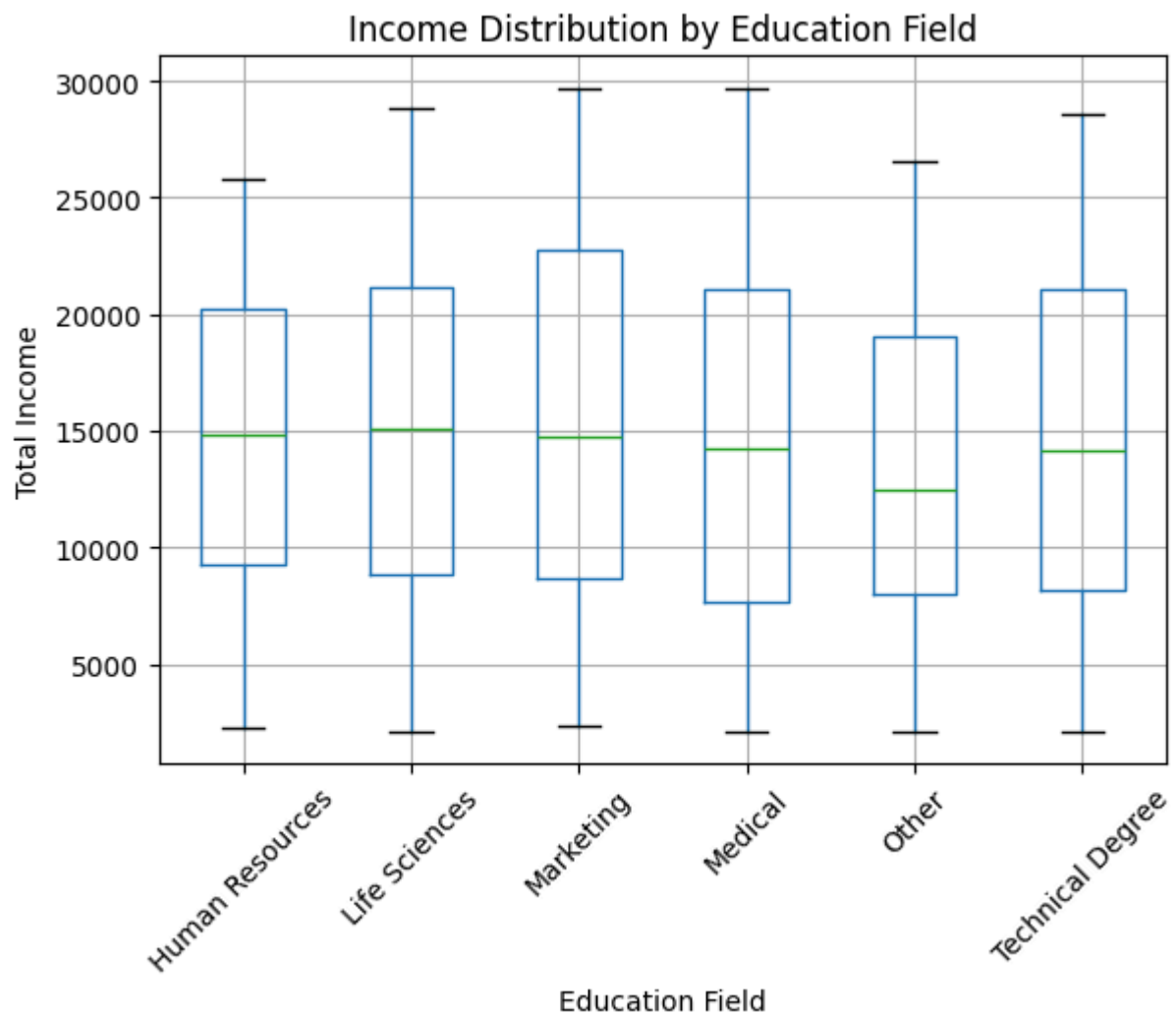
## How does the salary vary among graduates from different fields of study?

```
In [138... df['EducationField'].unique()
```

```
Out[138... array(['Life Sciences', 'Other', 'Medical', 'Marketing',
        'Technical Degree', 'Human Resources'], dtype=object)
```

```
In [139... # Boxplot for salary distribution by education field
plt.figure(figsize=(12, 5))
df.boxplot(column='TotalIncome', by='EducationField', grid=True, rot=45)
plt.title("Income Distribution by Education Field")
plt.xlabel("Education Field")
plt.ylabel("Total Income")
plt.suptitle("")
plt.show()
```

<Figure size 1200x500 with 0 Axes>



```
In [140]: max_income_by_EducateField = df.groupby('EducationField')['TotalIncome'].max()
max_income_by_EducateField
```

```
Out[140]: EducationField
Human Resources      25811.0
Life Sciences        28824.4
Marketing            29654.9
Medical              29696.7
Other                26537.0
Technical Degree     28547.2
Name: TotalIncome, dtype: float64
```

```
In [141]: min_income_by_EducateField = df.groupby('EducationField')['TotalIncome'].min()
min_income_by_EducateField
```

```
Out[141]: EducationField
Human Resources      2243.0
Life Sciences        2125.0
Marketing            2350.7
Medical              2094.0
Other                2112.0
Technical Degree     2125.0
Name: TotalIncome, dtype: float64
```

```
In [142]: avg_income_by_EducateField = df.groupby('EducationField')['TotalIncome'].mean()
avg_income_by_EducateField
```

```
Out[142... EducationField
Human Resources    14810.740741
Life Sciences      14889.631683
Marketing          15484.637736
Medical            14581.010991
Other              13545.251220
Technical Degree   14584.415909
Name: TotalIncome, dtype: float64
```

### How does the salary vary among graduates from different fields of study?

**DIAGNOS** : The box plot indicates that when comparing the **highest total incomes**, the **Marketing** and **Medical** fields have similar earnings, both ranking at the top. When looking at the **lowest total incomes**, all fields appear to have similar distributions. However, when comparing **average incomes**, most fields show very similar values, except for the **Other** category, which has a noticeably lower average than the rest. 📊

เนื่องจากโคบายาชิซังจะเป็นคนตรวจโค้ดของน้องๆ

ขอให้น้องๆแสดงข้อมูลอย่างเข้าใจง่ายและแม่นยำ ไม่งั้นโคบายาชิซังจะทำการไล่น้องๆออกจากบริษัทนะครับ

