

Task 3:- Operations on Multi-Index

1. Task Description:

Work with multi-index DataFrames and perform operations on different levels.

2. Task Output Screenshot:

```
# We will do multiIndex DataFrame using 'Country' and 'Item_Type'
data.set_index(['Country', 'Item_Type'], inplace=True)
print("\nMultiIndex DataFrame:\n", data)
```

MultiIndex DataFrame:

		Region \
Country	Item_Type	
Tuvalu	Baby Food	Australia and Oceania
Grenada	Cereal	Central America and the Caribbean
Russia	Office Supplies	Europe
Sao Tome and Principe	Fruits	Sub-Saharan Africa
Rwanda	Office Supplies	Sub-Saharan Africa
...		...
Mali	Clothes	Sub-Saharan Africa
Malaysia	Fruits	Asia
Sierra Leone	Vegetables	Sub-Saharan Africa
Mexico	Personal Care	North America
Mozambique	Household	Sub-Saharan Africa

		Sales_Channel	Order_Priority \
Country	Item_Type		
Tuvalu	Baby Food	Offline	H
Grenada	Cereal	Online	C
Russia	Office Supplies	Offline	L
Sao Tome and Principe	Fruits	Online	C
Rwanda	Office Supplies	Offline	L
...	
Mali	Clothes	Online	M
Malaysia	Fruits	Offline	L
Sierra Leone	Vegetables	Offline	C
Mexico	Personal Care	Offline	M
Mozambique	Household	Offline	L

		Ship_Date	Unit_Cost	Total_Revenue \
Country	Item_Type			
Tuvalu	Baby Food	27/06/2010	159.42	2533654.00
Grenada	Cereal	15/09/2012	117.11	576782.00
Russia	Office Supplies	05/08/2014	524.96	1158502.59
Sao Tome and Principe	Fruits	07/05/2014	6.92	75591.66
Rwanda	Office Supplies	02/06/2013	524.96	3296425.02
...	
Mali	Clothes	09/03/2011	35.84	97048.64
Malaysia	Fruits	28/12/2011	6.92	58471.11
Sierra Leone	Vegetables	29/06/2016	98.93	228779.10
Mexico	Personal Care	08/08/2015	56.67	471336.91
Mozambique	Household	15/02/2012	502.54	3506605.09

		Total_Profit
Country	Item_Type	
Tuvalu	Baby Food	951410.50
Grenada	Cereal	248406.36
Russia	Office Supplies	224598.75
Sao Tome and Principe	Fruits	19525.82
Rwanda	Office Supplies	639877.50
...		...
Mali	Clothes	65214.72
Malaysia	Fruits	15103.47
Sierra Leone	Vegetables	93748.05
Mexico	Personal Care	144521.02
Mozambique	Household	889472.91

[100 rows x 7 columns]

```
print("\nData for Country 'Mexico':\n", data.loc['Mexico'])
```

Data for Country 'Mexico':

	Region	Sales_Channel	Order_Priority	Ship_Date
Item_Type				
Household	North America	Offline	C	12/12/2014
Personal Care	North America	Offline	L	28/03/2012
Personal Care	North America	Offline	M	08/08/2015

	Unit_Cost	Total_Revenue	Total_Profit
Item_Type			
Household	582.54	4647149.58	1152486.42
Personal Care	56.67	524878.06	168935.32
Personal Care	56.67	471336.91	144521.02

```
print("\nData for Country 'Mexico' and Item Type 'Personal Care':\n", data.loc['Mexico', 'Personal Care'])
```

Data for Country 'Mexico' and Item Type 'Personal Care':

		Region	Sales_Channel	Order_Priority	Ship_Date
Country	Item_Type				
Mexico	Personal Care	North America	Offline	L	28/03/2012
	Personal Care	North America	Offline	M	08/08/2015

		Unit_Cost	Total_Revenue	Total_Profit
Country	Item_Type			
Mexico	Personal Care	56.67	524878.06	168935.32
	Personal Care	56.67	471336.91	144521.02

```
#Calculate the mean of Total Revenue for each country
mean_revenue_by_country_item = data.groupby(level=[0, 1])['Total_Revenue'].mean()
print("\nMean Total Revenue by Country and Item Type:\n", mean_revenue_by_country_item)
```

Mean Total Revenue by Country and Item Type:

Country	Item_Type	
Albania	Clothes	247956.32
Angola	Household	2798846.49
Australia	Beverages	445588.05
	Cereal	148287.40
	Office Supplies	1984138.04
	...	
Turkmenistan	Household	2559474.10
	Office Supplies	3262562.10
Tuvalu	Baby Food	2533654.00
United Kingdom	Household	188452.14
Zambia	Snacks	623289.30

Name: Total_Revenue, Length: 97, dtype: float64

```
# Swapping Levels
swapped_sales = data.swaplevel()
print("\nSwapped MultiIndex DataFrame:\n", swapped_sales)
```

Swapped MultiIndex DataFrame:

Item_Type	Country	Region \
Baby Food	Tuvalu	Australia and Oceania
Cereal	Grenada	Central America and the Caribbean
Office Supplies	Russia	Europe
Fruits	Sao Tome and Principe	Sub-Saharan Africa
Office Supplies	Rwanda	Sub-Saharan Africa
...		...
Clothes	Mali	Sub-Saharan Africa
Fruits	Malaysia	Asia
Vegetables	Sierra Leone	Sub-Saharan Africa
Personal Care	Mexico	North America
Household	Mozambique	Sub-Saharan Africa

Item_Type	Country	Sales_Channel	Order_Priority \
Baby Food	Tuvalu	Offline	H
Cereal	Grenada	Online	C
Office Supplies	Russia	Offline	L
Fruits	Sao Tome and Principe	Online	C
Office Supplies	Rwanda	Offline	L
...	
Clothes	Mali	Online	M
Fruits	Malaysia	Offline	L
Vegetables	Sierra Leone	Offline	C
Personal Care	Mexico	Offline	M
Household	Mozambique	Offline	L

Item_Type	Country	Ship_Date	Unit_Cost	Total_Revenue \
Baby Food	Tuvalu	27/06/2010	159.42	2533654.00
Cereal	Grenada	15/09/2012	117.11	576782.88
Office Supplies	Russia	05/08/2014	524.96	1158582.59
Fruits	Sao Tome and Principe	07/05/2014	6.92	75591.66
Office Supplies	Rwanda	02/06/2013	524.96	3296425.02
...	
Clothes	Mali	09/03/2011	35.84	97848.64
Fruits	Malaysia	28/12/2011	6.92	58471.11
Vegetables	Sierra Leone	29/06/2016	98.93	228779.10
Personal Care	Mexico	08/08/2015	56.67	471336.91
Household	Mozambique	15/02/2012	582.54	3586685.09

Item_Type	Country	Total_Profit
Baby Food	Tuvalu	951410.58
Cereal	Grenada	248486.36
Office Supplies	Russia	224598.75
Fruits	Sao Tome and Principe	19525.82
Office Supplies	Rwanda	639877.58
...		...
Clothes	Mali	65214.72
Fruits	Malaysia	15183.47
Vegetables	Sierra Leone	93748.05
Personal Care	Mexico	144521.02
Household	Mozambique	889472.91

[100 rows x 7 columns]

```
#Sorting the MultiIndex
sorted_sales = data.sort_index()
print("\nSorted MultiIndex DataFrame:\n", sorted_sales)
```

Sorted MultiIndex DataFrame:

Country	Item_Type	Region	Sales_Channel	\
Albania	Clothes	Europe	Online	
Angola	Household	Sub-Saharan Africa	Offline	
Australia	Beverages	Australia and Oceania	Offline	
	Cereal	Australia and Oceania	Offline	
	Office Supplies	Australia and Oceania	Online	
...		
Turkmenistan	Household	Asia	Offline	
	Office Supplies	Asia	Online	
Tuvalu	Baby Food	Australia and Oceania	Offline	
United Kingdom	Household	Europe	Online	
Zambia	Snacks	Sub-Saharan Africa	Online	

Country	Item_Type	Order_Priority	Ship_Date	Unit_Cost	\
Albania	Clothes	C	18/03/2010	35.84	
Angola	Household	M	27/04/2011	502.54	
Australia	Beverages	H	07/11/2014	31.79	
	Cereal	H	07/02/2013	117.11	
	Office Supplies	C	25/11/2015	524.96	
...		
Turkmenistan	Household	L	20/01/2011	502.54	
	Office Supplies	M	20/05/2013	524.96	
Tuvalu	Baby Food	H	27/06/2010	159.42	
United Kingdom	Household	L	14/02/2012	502.54	
Zambia	Snacks	L	01/05/2011	97.44	

Country	Item_Type	Total_Revenue	Total_Profit
Albania	Clothes	247956.32	166635.36
Angola	Household	2798046.49	693911.51
Australia	Beverages	445588.05	147831.74
	Cereal	140287.40	60418.38
	Office Supplies	1984138.04	369155.00
...	
Turkmenistan	Household	2559474.10	634745.90
	Office Supplies	3262562.10	632512.50
Tuvalu	Baby Food	2533654.00	951410.50
United Kingdom	Household	188452.14	46735.86
Zambia	Snacks	623289.30	225246.90

[100 rows x 7 columns]

```
#Let's Calculate the total profit for each country
total_profit_by_country = data.groupby(level=0)['Total_Profit'].sum()
print("\nTotal Profit by Country:\n", total_profit_by_country)
```

Total Profit by Country:

Country	
Albania	166635.36
Angola	693911.51
Australia	576605.12
Austria	495807.89
Azerbaijan	1512926.83
...	...
The Gambia	1385883.27
Turkmenistan	1267258.40
Tuvalu	951410.50
United Kingdom	46735.86
Zambia	225246.90

Name: Total_Profit, Length: 76, dtype: float64

```
#Let's Calculate the total revenue for each item type across all countries
total_revenue_by_item_type = data.groupby(level=1)['Total_Revenue'].sum()
print("\nTotal Revenue by Item Type:\n", total_revenue_by_item_type)
```

```
Total Revenue by Item Type:
Item_Type
Baby Food      18358327.68
Beverages      2698794.68
Cereal         5322898.98
Clothes        7787292.88
Cosmetics      36681589.68
Fruits         466481.34
Household      29889712.29
Meat           4583675.75
Office Supplies 38585388.87
Personal Care  3988984.84
Snacks         2888733.46
Vegetables     3889857.86
Name: Total_Revenue, dtype: float64
```

3. Algorithm Used In Task:

- The concept of MultiIndex is used, which allows for more complex and hierarchical data representations, with multiple levels of indexing (e.g., Country and Item_Type).
- Used groupby() to aggregate data based on the multi-index levels (e.g., calculating mean revenue or total profit by country or item type).
- Methods like here used are swaplevel() to swap index levels and sort_index() to reorder the multi-index based on different criteria.
- Operations like loc is used for selecting specific data, and sum() or mean() for aggregating numeric values.