```
In [4]: import pandas as pd

# Load the Excel file
df = pd.read_excel(r"C:\Users\ASUS\Downloads\Synthetic Agent Data (1).xlsx")

# Display the first few rows to understand the structure
df.head()
```

Out[4]:		Unnamed: 0	Agent 0	Agent 1	Agent 2	Agent 3	Agent 4	Agent 5	Agent 6	Agent 7	Agent 8	•••	Age 11
	0	0	1	2	1	0	1	1	0	0	0		
	1	1	9	10	8	2	15	7	0	2	2		
	2	2	9	7	11	3	10	7	0	1	3		
	3	3	11	14	13	0	8	9	0	0	3		
	4	4	11	13	7	1	7	8	1	0	4		

5 rows × 1201 columns

```
In [5]: #Dropping the 'unnamed column: 0' and removing the column from the original
df.drop(columns = 'Unnamed: 0', axis=1, inplace=True)
```

In [6]: **df**

Out[6]:		Agent 0	Agent 1	Agent 2	Agent 3	Agent 4	Agent 5	Agent 6	Agent 7	Agent 8	Agent 9	•••	Agent 1190
	0	1	2	1	0	1	1	0	0	0	2	•••	0
	1	9	10	8	2	15	7	0	2	2	11		0
	2	9	7	11	3	10	7	0	1	3	11		3
	3	11	14	13	0	8	9	0	0	3	14		1
	4	11	13	7	1	7	8	1	0	4	13		2
	•••	•••			•••		•••		•••	•••			
	79	8	8	7	1	10	8	0	1	2	8		0
	80	7	11	7	1	7	7	3	2	3	8		0
	81	11	7	9	1	13	10	0	1	3	8		1
	82	9	9	11	0	7	7	0	0	2	7		2
	83	2	2	2	0	2	2	0	0	0	1		0

84 rows × 1200 columns

In [10]: #Question 1 #What are the mean, median, maximum, and minimum number of tasks comple
#What is the standard deviation of their number of tasks completed?
#Please also calculate the same statistics (mean, median, max, and minimum) for the

Calculate statistics for each agent
agent_stats = df.describe().transpose()
agent_stats

Out[10]:		count	mean	std	min	25%	50%	75%	max
	Agent 0	84.0	6.940476	3.816108	1.0	2.0	7.0	10.00	14.0
	Agent 1	84.0	6.702381	3.867628	1.0	2.0	7.0	9.00	15.0
	Agent 2	84.0	7.000000	4.032996	1.0	2.0	7.5	9.00	15.0
	Agent 3	84.0	0.535714	0.870486	0.0	0.0	0.0	1.00	3.0
	Agent 4	84.0	7.297619	4.236331	1.0	2.0	8.0	10.00	15.0
	•••		•••						
	Agent 1195	84.0	0.309524	0.559177	0.0	0.0	0.0	1.00	2.0
	Agent 1196	84.0	2.083333	1.716036	0.0	0.0	2.0	3.00	6.0
	Agent 1197	84.0	1.714286	1.923673	0.0	0.0	1.0	3.00	8.0
	Agent 1198	84.0	7.285714	4.249937	1.0	2.0	7.0	10.25	15.0
	Agent 1199	84.0	0.595238	0.879746	0.0	0.0	0.0	1.00	3.0

1200 rows × 8 columns

```
In [12]: # Calculate statistics for all agents combined
         combined_stats = {
             'mean': df.values.mean(),
             'median': pd.Series(df.values.flatten()).median(),
             'max': df.values.max(),
             'min': df.values.min(),
             'std': df.values.std()
         combined stats
Out[12]: {'mean': 3.5428571428571427,
           'median': 1.0,
           'max': 24,
           'min': 0,
           'std': 4.230781543835055}
In [14]: #Question 2: Not all agents are the same, some are driven by decidedly different fa
         #From the data, what can you say about the different types of agents?
         from sklearn.cluster import KMeans
         # Prepare data for clustering (e.g., use mean and std of tasks completed)
         #X = df.describe().transpose()[['mean', 'std']]
         # Perform K-means clustering
         kmeans = KMeans(n_clusters=3)
         agent_stats['Cluster'] = kmeans.fit_predict(agent_stats)
```

```
# Add cluster labels to the data
print(agent_stats)
```

```
C:\Users\ASUS\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: FutureWarn
ing: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the val
ue of `n_init` explicitly to suppress the warning
 warnings.warn(
```

C:\Users\ASUS\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1382: UserWarni ng: KMeans is known to have a memory leak on Windows with MKL, when there are less c hunks than available threads. You can avoid it by setting the environment variable O MP_NUM_THREADS=5.

warnings.warn(

```
25% 50%
                                                    75%
           count
                     mean
                                std min
                                                          max Cluster
Agent 0
            84.0 6.940476 3.816108 1.0
                                         2.0 7.0 10.00 14.0
                                                                    1
            84.0 6.702381 3.867628 1.0
                                                  9.00 15.0
Agent 1
                                         2.0 7.0
                                                                    1
Agent 2
            84.0 7.000000 4.032996 1.0
                                         2.0 7.5
                                                   9.00 15.0
                                                                    1
Agent 3
            84.0 0.535714 0.870486 0.0
                                         0.0 0.0
                                                   1.00
                                                          3.0
                                                                    0
Agent 4
            84.0 7.297619 4.236331 1.0
                                         2.0 8.0 10.00 15.0
                                                                     1
             . . .
                      . . .
                                . . .
                                    . . .
                                         . . .
                                             . . .
                                                    . . .
                                                                   . . .
Agent 1195
            84.0 0.309524 0.559177
                                    0.0
                                         0.0 0.0
                                                   1.00
                                                          2.0
                                                                    0
                                                   3.00
                                                                    2
Agent 1196
            84.0 2.083333 1.716036 0.0
                                         0.0 2.0
                                                          6.0
Agent 1197
            84.0 1.714286 1.923673 0.0
                                         0.0 1.0
                                                   3.00
                                                          8.0
                                                                    2
            84.0 7.285714 4.249937 1.0
                                         2.0 7.0 10.25 15.0
Agent 1198
Agent 1199
            84.0 0.595238 0.879746 0.0 0.0 0.0
                                                  1.00
                                                          3.0
```

[1200 rows x 9 columns]

```
In [18]: # Cluster Summary
         cluster_summary = agent_stats.groupby('Cluster').mean()
         print(cluster_summary)
```

	count	mean	std	min	25%	50%	75%	\
Cluster								
0	84.0	0.515381	0.811575	0.000000	0.000000	0.072519	1.017653	
1	84.0	7.093262	4.183311	0.965164	1.929816	7.404713	9.887807	
2	84.0	2.765198	2.606393	0.000000	0.000000	2.473404	4.498670	

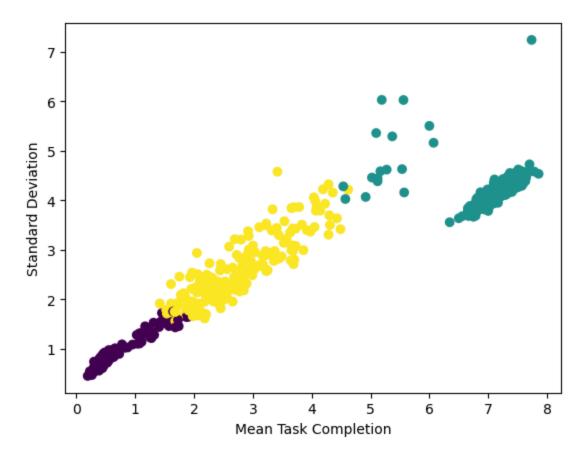
max

Cluster 0

3.074427 14.948770

8.952128

```
In [16]: #Use matplotlib to plot the clusters and interpret the different agent personas
         import matplotlib.pyplot as plt
         plt.scatter(agent_stats['mean'], agent_stats['std'], c=agent_stats['Cluster'])
         plt.xlabel('Mean Task Completion')
         plt.ylabel('Standard Deviation')
         plt.show()
```



In []: #For each cluster, summarize the characteristics as their persona:
 #High Performers: Cluster 2 with High mean, low standard deviation indicates c
#Inconsistent Performers: Cluster 1 with High standard deviation might represen
#Low Performers: Cluster 0 with Low mean and low standard deviation might indic

#What contributed to their different persona?
#Experience and Skill Level: More experienced agents often perform better due to
#Work Environment: The work environment, including the quality of tools, support
 #A supportive and well-equipped environment can boost product
#Personal Factors: Individual differences such as health, stress levels, and per
#Incentives and Rewards: Agents who are aware of and are motivated by the incent
#incentives align with their goals and aspirations.

In []: #Question 3: We would like to design a compensation structure for the agents that s
#but are not quite the best yet. Really, we're trying to motivate them
#We'd like the total average compensation for a task to be at most 300K
#We'd like every task to have a minimum payment of at least 175KSH so a

#Please design an incentive structure for agents in the form of a per task rate and
#You can create a single flat bonus, multiple tiers of incentives, or anything that
#Your solution will be judged based on how well it met the financial criteria, your
#and the ease of which the incentives can be explained to agents.

```
In [32]: # Define base rate and tiers

BASE_RATE = 250  # Base payment per task
BONUS_TIERS = {  # Weekly bonus structure
    'Tier 1': (20, 30, 500),  # Range: 20-30 tasks, Bonus: 500 KSH
    'Tier 2': (31, 40, 1000),  # Range: 31-40 tasks, Bonus: 1000 KSH
```

```
'Tier 3': (41, float('inf'), 2000) # 41+ tasks, Bonus: 2000 KSH
}
def calculate_bonus(tasks_completed):
   This function returns the bonus based on tasks completed.
   for tier, (min_tasks, max_tasks, bonus) in BONUS_TIERS.items():
        if min tasks <= tasks completed <= max tasks:</pre>
            return bonus
   return 0 # No bonus if tasks are below 20
def calculate total compensation(tasks completed):
   This function calculates the total compensation including base pay and bonus.
   # Base pay is simply the number of tasks completed times the base rate
   base_pay = tasks_completed * BASE_RATE
   # Calculate bonus
   bonus = calculate_bonus(tasks_completed)
   # Total compensation is base pay + bonus
   total_compensation = base_pay + bonus
   return total compensation
def calculate_average_pay_per_task(tasks_completed):
   This function calculates the average pay per task.
   if tasks completed == 0:
        return 0 # Avoid division by zero
   # Total compensation
   total_compensation = calculate_total_compensation(tasks_completed)
   # Average pay per task
   average_pay_per_task = total_compensation / tasks_completed
   return average_pay_per_task
# Example usage:
tasks_completed = int(input("Enter the number of tasks completed: "))
# Calculate total compensation and average pay per task
total_compensation = calculate_total_compensation(tasks_completed)
average_pay = calculate_average_pay_per_task(tasks_completed)
print(f"Total Compensation: {total compensation} KSH")
print(f"Average Pay Per Task: {average_pay:.2f} KSH")
```

Total Compensation: 5500 KSH Average Pay Per Task: 275.00 KSH

```
In [ ]:
```

In []: