

```
# Latihan 1
# Import library numpy untuk operasi fungsi aritmatika
import numpy as np

# import library pandas untuk operasi dataframe
import pandas as pd

# Import library matplotlib dan seaborn untuk visualisasi
import matplotlib
import matplotlib.pyplot as plt
from matplotlib import colors
import seaborn as sns

# Import library Axes3D untuk vizualisasi 3 Dimensi
from mpl_toolkits.mplot3d import Axes3D

# import library datetime untuk operasi yang berhubungan dengan waktu.
import datetime

# import library Label encoder untuk mengubah setiap nilai dalam kolom
menjadi angka yang berurutan / numeric
from sklearn.preprocessing import LabelEncoder

# import library StandardScaler untuk menskalakan nilai kolom jika
terdapat perbedaan skala, StandardScaler berfungsi menghilangkan mean
(terpusat pada 0) dan menskalakan ke variansi (deviasi standar = 1),
dengan asumsi data terdistribusi normal (gauss) untuk semua fitur
from sklearn.preprocessing import StandardScaler

# import library PCA adalah prosedur statistik yang mengekstrak fitur-
fitur terpenting dari suatu dataset
from sklearn.decomposition import PCA

# import library KElbowVisualizer untuk mengimplementasikan metode
"elbow/siku" untuk data scientist memilih jumlah cluster yang optimal
dengan menyesuaikan model dengan rentang nilai untuk K
from yellowbrick.cluster import KElbowVisualizer

# import library KMeans metode adalah teknik unsupervised machine
learning yang digunakan untuk mengidentifikasi kelompok objek data
dalam kumpulan data
from sklearn.cluster import KMeans

# import library AgglomerativeClustering untuk melakukan pengelompokan
data menggunakan bottom-up manner
from sklearn.cluster import AgglomerativeClustering

# import library metrics untuk mengimplementasikan fungsi yang menilai
```

*kesalahan prediksi untuk tujuan tertentu*

```
from sklearn import metrics
```

*# me-non aktifkan peringatan pada python*

```
import warnings
```

```
import sys
```

```
if not sys.warnoptions:
```

```
    warnings.simplefilter("ignore")
```

*# mendefinisikan nilai acak*

```
np.random.seed(42)
```

*# Latihan 2*

*#Load the dataset dan tampilkan data nya*

```
data = pd.read_csv('marketing_campaign.csv', sep="\t")
```

```
print("Number of datapoints:", len(data))
```

```
data.head()
```

Number of datapoints: 2240

	ID	Year_Birth	Education	...	Z_CostContact	Z_Revenue
Response						
0	5524	1957	Graduation	...	3	11
1						
1	2174	1954	Graduation	...	3	11
0						
2	4141	1965	Graduation	...	3	11
0						
3	6182	1984	Graduation	...	3	11
0						
4	5324	1981	PhD	...	3	11
0						

[5 rows x 29 columns]

*# Latihan 3*

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 2240 entries, 0 to 2239
```

```
Data columns (total 29 columns):
```

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	ID	2240 non-null	int64
1	Year_Birth	2240 non-null	int64
2	Education	2240 non-null	object
3	Marital_Status	2240 non-null	object
4	Income	2216 non-null	float64
5	Kidhome	2240 non-null	int64
6	Teenhome	2240 non-null	int64
7	Dt_Customer	2240 non-null	object

```

8   Recency                2240 non-null   int64
9   MntWines               2240 non-null   int64
10  MntFruits              2240 non-null   int64
11  MntMeatProducts        2240 non-null   int64
12  MntFishProducts        2240 non-null   int64
13  MntSweetProducts       2240 non-null   int64
14  MntGoldProds           2240 non-null   int64
15  NumDealsPurchases      2240 non-null   int64
16  NumWebPurchases        2240 non-null   int64
17  NumCatalogPurchases   2240 non-null   int64
18  NumStorePurchases      2240 non-null   int64
19  NumWebVisitsMonth      2240 non-null   int64
20  AcceptedCmp3           2240 non-null   int64
21  AcceptedCmp4           2240 non-null   int64
22  AcceptedCmp5           2240 non-null   int64
23  AcceptedCmp1           2240 non-null   int64
24  AcceptedCmp2           2240 non-null   int64
25  Complain               2240 non-null   int64
26  Z_CostContact           2240 non-null   int64
27  Z_Revenue              2240 non-null   int64
28  Response               2240 non-null   int64
dtypes: float64(1), int64(25), object(3)
memory usage: 507.6+ KB

```

```
# menghapus missing values
```

```

data = data.dropna()
print("Jumlah data setelah menghapus baris dengan nilai yang hilang
adalah:", len(data))

```

Jumlah data setelah menghapus baris dengan nilai yang hilang adalah:  
2216

```

data["Dt_Customer"] = pd.to_datetime(data["Dt_Customer"])
dates = []
for i in data["Dt_Customer"]:
    i = i.date()
    dates.append(i)

```

```
# Tanggal pelanggan terbaru dan terlama yang tercatat
print("Tanggal pendaftaran pelanggan terbaru dalam
catatan:", max(dates))
print("Tanggal pendaftaran pelanggan terlama dalam
catatan:", min(dates))

```

Tanggal pendaftaran pelanggan terbaru dalam catatan: 2014-12-06  
Tanggal pendaftaran pelanggan terlama dalam catatan: 2012-01-08

```
# Membuat fitur "Customer_For"
```

```

days = []
d1 = max(dates) # membawanya menjadi pelanggan terbaru
for i in dates:

```

```

    delta = d1 - i
    days.append(delta)
data["Customer_For"] = days
data["Customer_For"] = pd.to_numeric(data["Customer_For"],
errors="coerce")

```

```

print("Total kategori dalam fitur Marital_Status:\n\n",
data["Marital_Status"].value_counts(), "\n")
print("Total kategori dalam fitur Education:\n\n",
data["Education"].value_counts())

```

Total kategori dalam fitur Marital\_Status:

```

Married      857
Together     573
Single       471
Divorced     232
Widow        76
Alone         3
YOLO         2
Absurd        2
Name: Marital_Status, dtype: int64

```

Total kategori dalam fitur Education:

```

Graduation   1116
PhD           481
Master       365
2n Cycle     200
Basic         54
Name: Education, dtype: int64

```

*# Latihan 4*  
*#Feature Engineering*

*# Usia pelanggan hari ini*  
data["Age"] = 2021-data["Year\_Birth"]

*# Total pengeluaran untuk berbagai macam item*  
data["Spent"] = data["MntWines"]+ data["MntFruits"]+  
data["MntMeatProducts"]+ data["MntFishProducts"]+  
data["MntSweetProducts"]+ data["MntGoldProds"]

*# situasi kehidupan dari status pernikahan "Alone"*  
data["Living\_With"]=data["Marital\_Status"].replace({"Married":"Partner",  
", "Together":"Partner", "Absurd":"Alone", "Widow":"Alone",  
"YOLO":"Alone", "Divorced":"Alone", "Single":"Alone",})

*# Fitur yang menunjukkan jumlah anak yang tinggal di rumah tangga*  
data["Children"]=data["Kidhome"]+data["Teenhome"]

```

# Fitur untuk total anggota dalam rumah tangga
data["Family_Size"] = data["Living_With"].replace({"Alone": 1,
"Partner":2})+ data["Children"]

# Fitur yang berkaitan dengan orang tua
data["Is_Parent"] = np.where(data.Children> 0, 1, 0)

# Segmentasi tingkat pendidikan dalam tiga kelompok
data["Education"]=data["Education"].replace({"Basic":"Undergraduate", "
2n Cycle":"Undergraduate", "Graduation":"Graduate",
"Master":"Postgraduate", "PhD":"Postgraduate"})

# Untuk kejelasan produk
data=data.rename(columns={"MntWines":
"Wines", "MntFruits":"Fruits", "MntMeatProducts":"Meat", "MntFishProducts
":"Fish", "MntSweetProducts":"Sweets", "MntGoldProds":"Gold"})

# Drop / Menjatuhkan beberapa fitur yang berlebihan / redundant
features
to_drop = ["Marital_Status", "Dt_Customer", "Z_CostContact",
"Z_Revenue", "Year_Birth", "ID"]
data = data.drop(to_drop, axis=1)

# melihat statistik data untuk data numeric
data.describe()

```

	Income	Kidhome	...	Family_Size	Is_Parent
count	2216.000000	2216.000000	...	2216.000000	2216.000000
mean	52247.251354	0.441787	...	2.592509	0.714350
std	25173.076661	0.536896	...	0.905722	0.451825
min	1730.000000	0.000000	...	1.000000	0.000000
25%	35303.000000	0.000000	...	2.000000	0.000000
50%	51381.500000	0.000000	...	3.000000	1.000000
75%	68522.000000	1.000000	...	3.000000	1.000000
max	666666.000000	2.000000	...	5.000000	1.000000

[8 rows x 28 columns]

```

# Latihan 5
#To plot some selected features
#Setting up colors preferences
sns.set(rc={"axes.facecolor":"#FFF9ED", "figure.facecolor":"#FFF9ED"})
pallet = ["#682F2F", "#9E726F", "#D6B2B1", "#B9C0C9", "#9F8A78",
"#F3AB60"]
cmap = colors.ListedColormap(["#682F2F", "#9E726F", "#D6B2B1",
"#B9C0C9", "#9F8A78", "#F3AB60"])

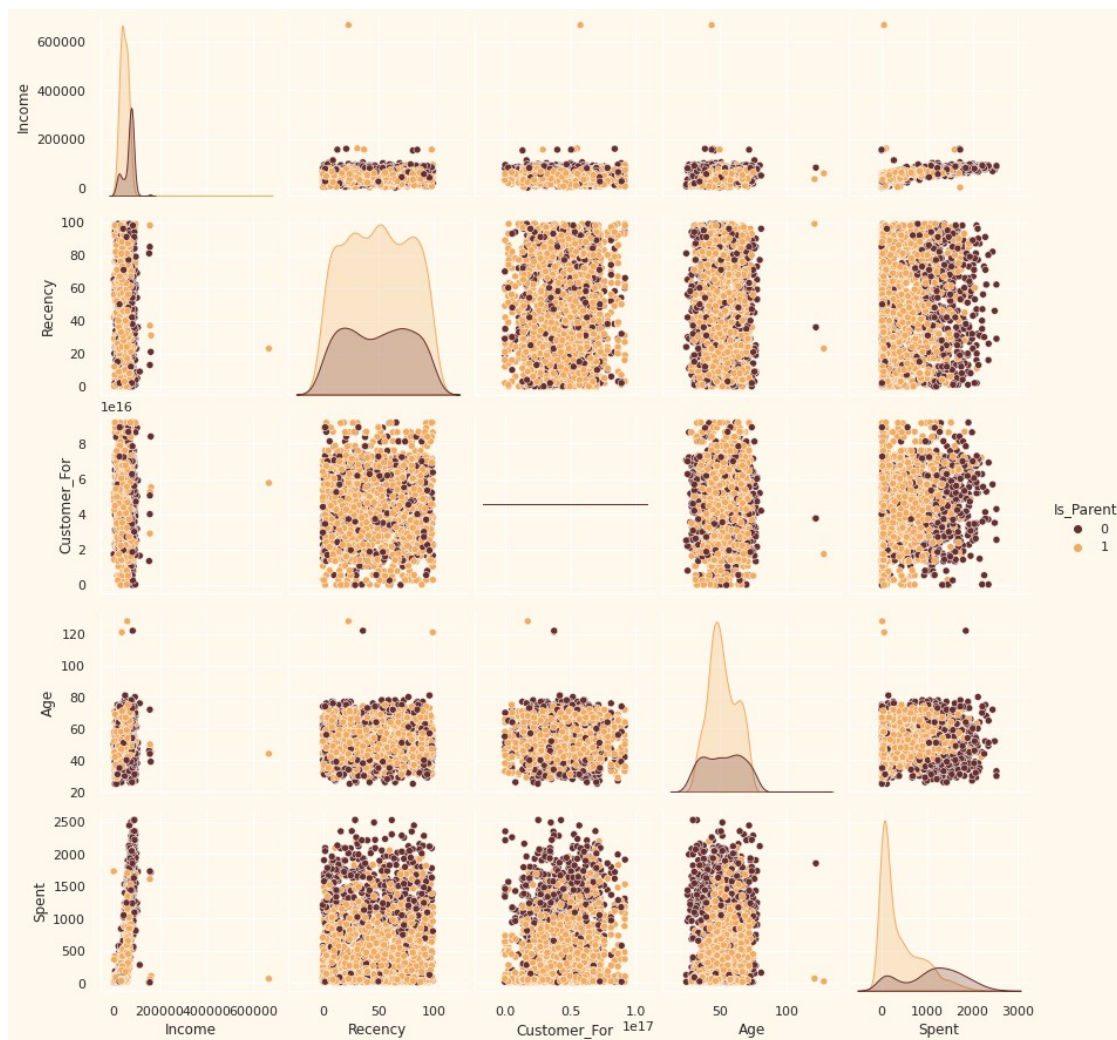
```

```
#Plotting fitur
To_Plot = [ "Income", "Recency", "Customer_For", "Age", "Spent",
            "Is_Parent"]
print("Plot Relatif Dari Beberapa Fitur Terpilih: Subset Data")
plt.figure()
sns.pairplot(data[To_Plot], hue= "Is_Parent",palette=
              ([ "#682F2F", "#F3AB60"])))

plt.show()
```

Plot Relatif Dari Beberapa Fitur Terpilih: Subset Data

<Figure size 432x288 with 0 Axes>



```
# Drop outlier dengan menetapkan batas pada Usia dan pendapatan.
data = data[(data["Age"]<90)]
data = data[(data["Income"]<600000)]
print("Jumlah total data setelah menghapus outlier adalah:",
      len(data))
```

Jumlah total data setelah menghapus outlier adalah: 2212

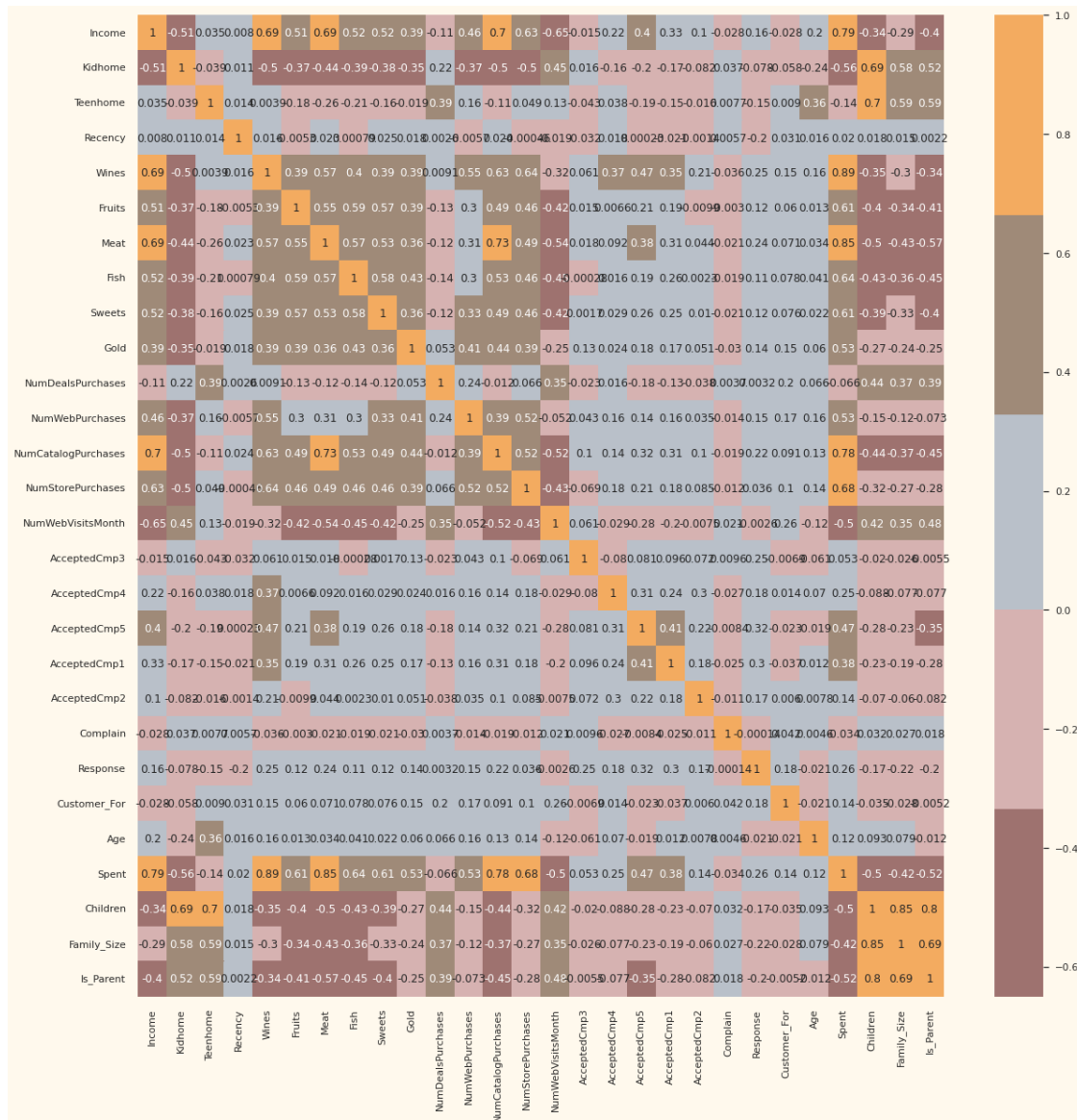
```
#correlation matrix
```

```
corrmat= data.corr()
```

```
plt.figure(figsize=(20,20))
```

```
sns.heatmap(corrmat,annot=True, cmap=cmap, center=0)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7effeb904510>
```



```
# Latihan 6
```

```
# Get List dari variabel categorical
```

```
s = (data.dtypes == 'object')
```

```
object_cols = list(s[s].index)
```

```
print("Variabel kategori dalam dataset:", object_cols)
```



Variabel kategori dalam dataset: ['Education', 'Living\_With']

```
# Label Encoding (dtypes: objek)
```

```
LE=LabelEncoder()
```

```
for i in object_cols:
```

```
    data[i]=data[[i]].apply(LE.fit_transform)
```

```
print("Semua fitur sekarang numerik")
```

Semua fitur sekarang numerik

```
# Membuat salinan data
```

```
ds = data.copy()
```

```
# membuat subset dataframe dengan menghapus fitur pada penawaran yang diterima (features on deals accepted) dan promosi (promotions)
```

```
cols_del = ['AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5',  
'AcceptedCmp1', 'AcceptedCmp2', 'Complain', 'Response']
```

```
ds = ds.drop(cols_del, axis=1)
```

```
#Scaling
```

```
scaler = StandardScaler()
```

```
scaler.fit(ds)
```

```
scaled_ds = pd.DataFrame(scaler.transform(ds), columns= ds.columns )
```

```
print("Semua fitur sekarang sudah diskalakan")
```

Semua fitur sekarang sudah diskalakan

```
# Data yang diskalakan untuk digunakan untuk mengurangi dimensi/reducing the dimensionality
```

```
print("Dataframe yang akan digunakan untuk pemodelan lebih lanjut:")
```

```
scaled_ds.head()
```

Dataframe yang akan digunakan untuk pemodelan lebih lanjut:

	Education	Income	Kidhome	...	Children	Family_Size	
Is_Parent							
0	-0.893586	0.287105	-0.822754	...	-1.264598	-1.758359	-
1	-0.893586	-0.260882	1.040021	...	1.404572	0.449070	
2	-0.893586	0.913196	-0.822754	...	-1.264598	-0.654644	-
3	-0.893586	-1.176114	1.040021	...	0.069987	0.449070	
4	0.571657	0.294307	1.040021	...	0.069987	0.449070	

[5 rows x 23 columns]

```
# Latihan 7
```

```
# Memulai PCA untuk mengurangi dimensi alias fitur menjadi 3
```



```
pca = PCA(n_components=3)
pca.fit(scaled_ds)
PCA_ds = pd.DataFrame(pca.transform(scaled_ds),
columns=["col1", "col2", "col3"])
PCA_ds.describe().T
```

	count	mean	std	...	50%	75%
max						
col1	2212.0	-8.612761e-17	2.878377	...	-0.780421	2.383290
7.444305						
col2	2212.0	2.559737e-17	1.706839	...	-0.158123	1.242289
6.142721						
col3	2212.0	4.449425e-17	1.221956	...	-0.022692	0.799895
6.611222						

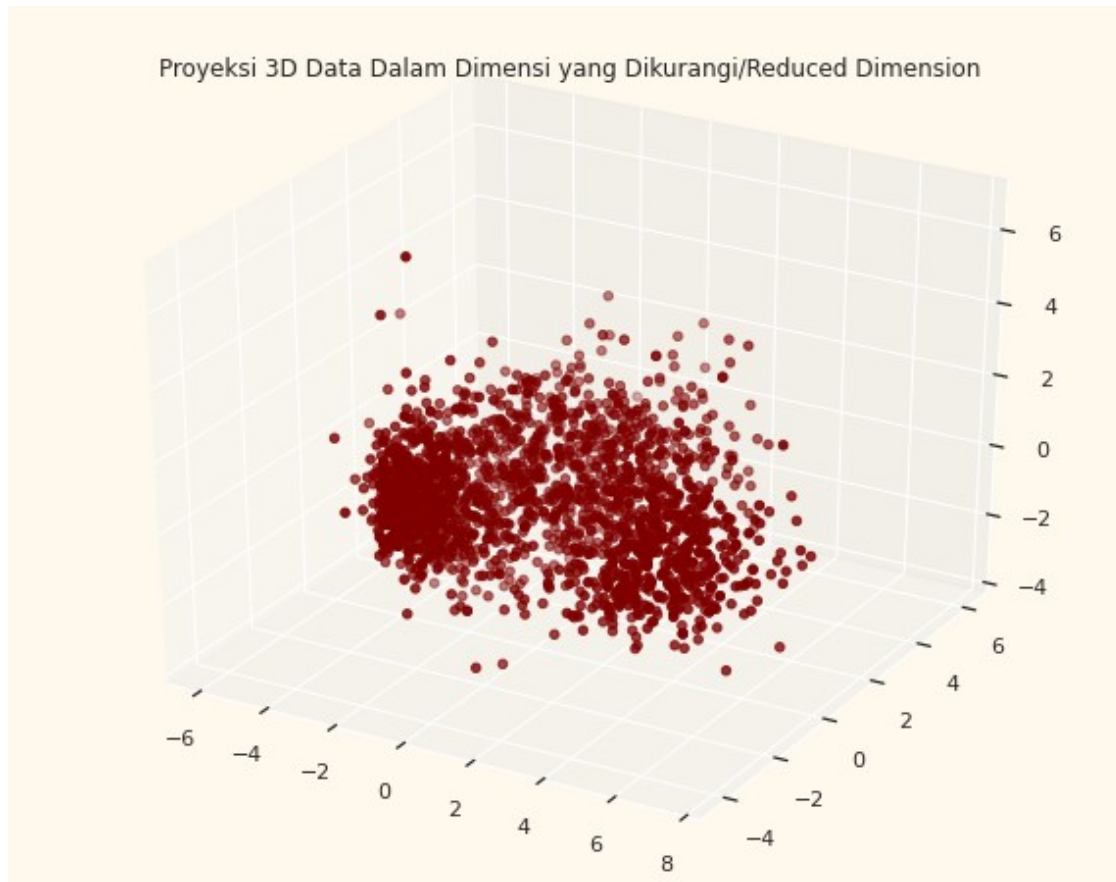
```
[3 rows x 8 columns]
```

```
# Proyeksi 3D Data Dalam Dimensi yang Dikurangi/Reduced Dimension
```

```
x =PCA_ds["col1"]
y =PCA_ds["col2"]
z =PCA_ds["col3"]
```

```
# plotting
```

```
fig = plt.figure(figsize=(10,8))
ax = fig.add_subplot(111, projection="3d")
ax.scatter(x,y,z, c="maroon", marker="o" )
ax.set_title("Proyeksi 3D Data Dalam Dimensi yang Dikurangi/Reduced Dimension")
plt.show()
```



*# Latihan 8*

*# Quick examination of elbow method to find numbers of clusters to make.*

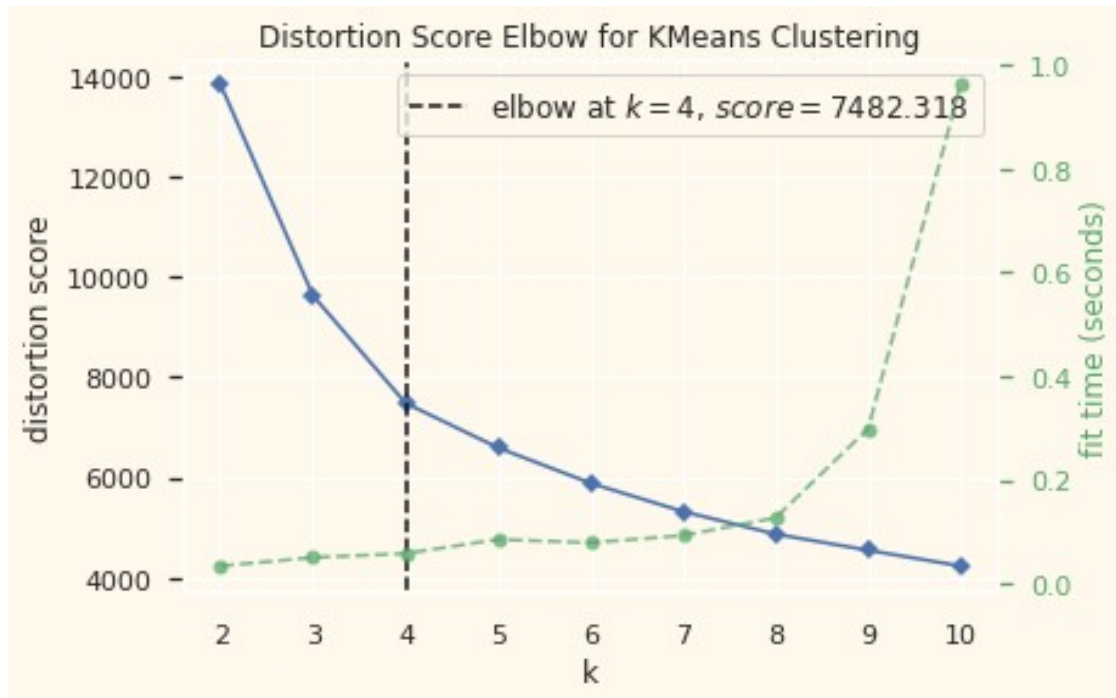
```
print('Metode Elbow untuk menentukan jumlah cluster yang akan  
dibentuk:')
```

```
Elbow_M = KElbowVisualizer(KMeans(), k=10)
```

```
Elbow_M.fit(PCA_ds)
```

```
Elbow_M.show()
```

Metode Elbow untuk menentukan jumlah cluster yang akan dibentuk:



```
<matplotlib.axes._subplots.AxesSubplot at 0x7effeb7a8650>
```

```
# Memulai model Agglomerative Clustering
```

```
AC = AgglomerativeClustering(n_clusters=4)
```

```
# fit model and predict clusters
```

```
yhat_AC = AC.fit_predict(PCA_ds)
```

```
PCA_ds["Clusters"] = yhat_AC
```

```
# Menambahkan fitur Cluster ke dataframe asli.
```

```
data["Clusters"] = yhat_AC
```

```
#Plotting clusters
```

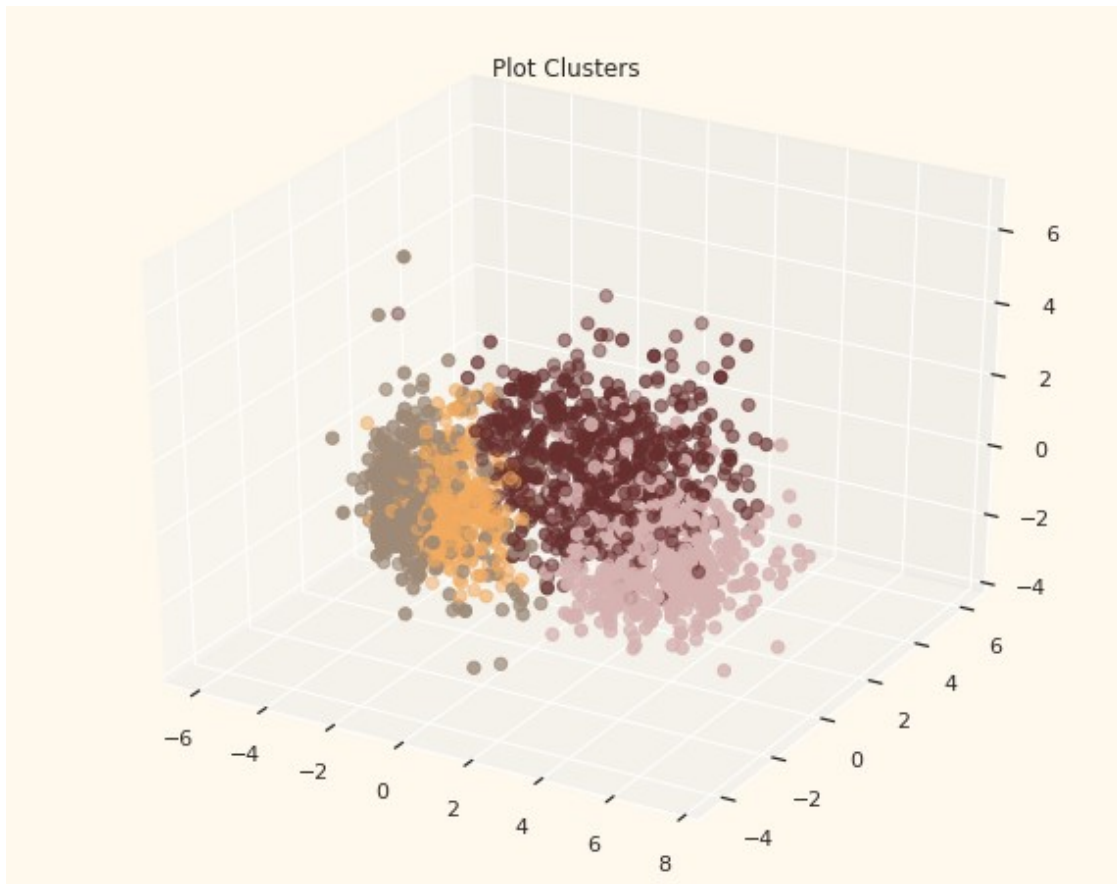
```
fig = plt.figure(figsize=(10,8))
```

```
ax = plt.subplot(111, projection='3d', label="bla")
```

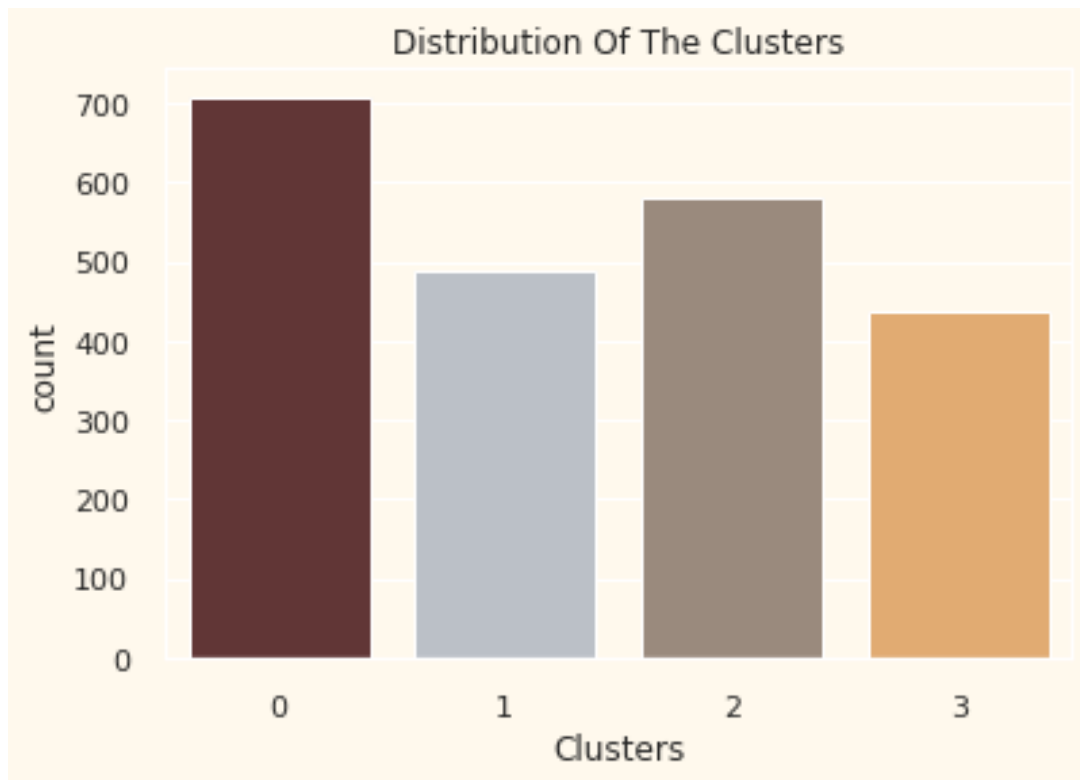
```
ax.scatter(x, y, z, s=40, c=PCA_ds["Clusters"], marker='o', cmap =  
cmap )
```

```
ax.set_title("Plot Clusters")
```

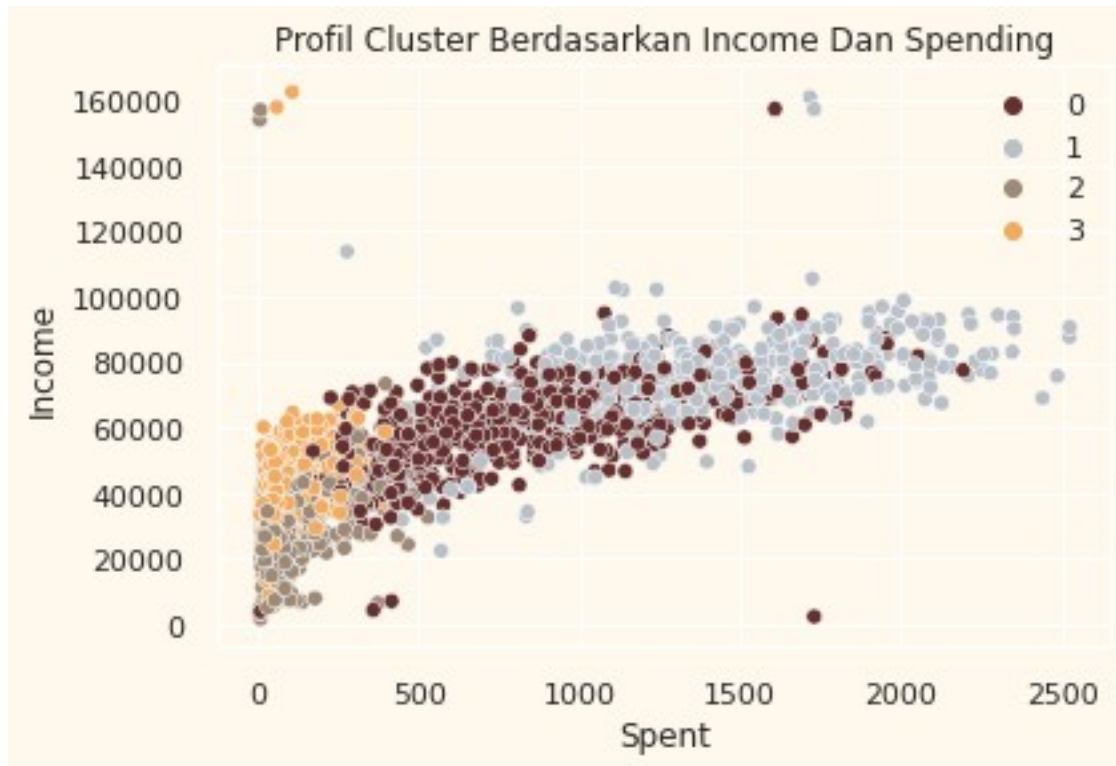
```
plt.show()
```



```
# Latihan 9
#Plotting countplot dari clusters
pal = ["#682F2F", "#B9C0C9", "#9F8A78", "#F3AB60"]
pl = sns.countplot(x=data["Clusters"], palette= pal)
pl.set_title("Distribution Of The Clusters")
plt.show()
```



```
pl = sns.scatterplot(data = data,x=data["Spent"],  
y=data["Income"],hue=data["Clusters"], palette= pal)  
pl.set_title("Profil Cluster Berdasarkan Income Dan Spending")  
plt.legend()  
plt.show()
```



```
plt.figure()
pl=sns.swarmplot(x=data["Clusters"], y=data["Spent"], color=
"#CBEDDD", alpha=0.5 )
pl=sns.boxenplot(x=data["Clusters"], y=data["Spent"], palette=pal)
plt.show()
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296:
UserWarning: 55.2% of the points cannot be placed; you may want to
decrease the size of the markers or use stripplot.
```

```
warnings.warn(msg, UserWarning)
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296:
UserWarning: 36.3% of the points cannot be placed; you may want to
decrease the size of the markers or use stripplot.
```

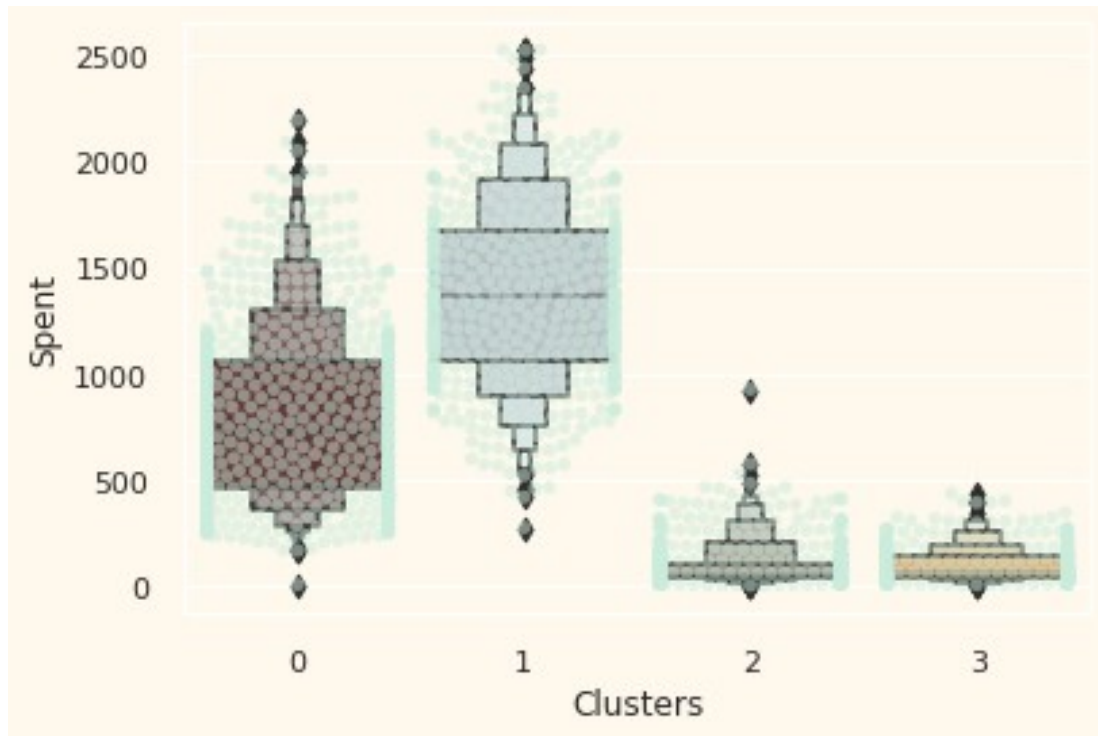
```
warnings.warn(msg, UserWarning)
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296:
UserWarning: 83.3% of the points cannot be placed; you may want to
decrease the size of the markers or use stripplot.
```

```
warnings.warn(msg, UserWarning)
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296:
UserWarning: 81.7% of the points cannot be placed; you may want to
decrease the size of the markers or use stripplot.
```

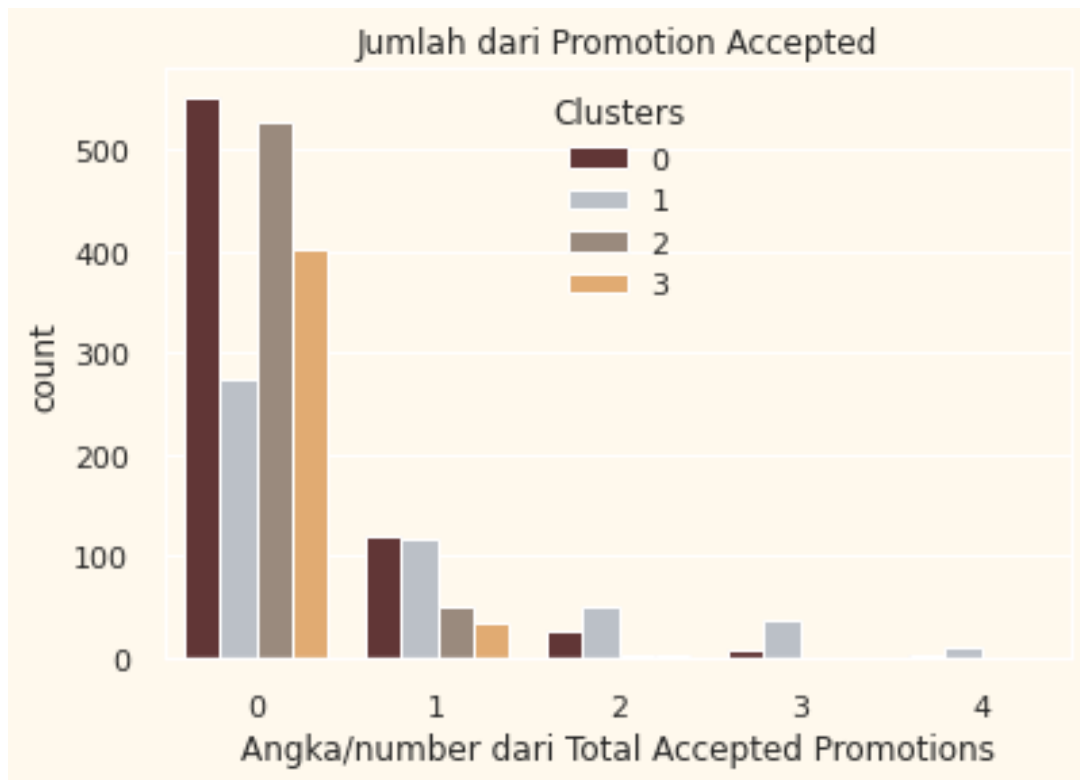
```
warnings.warn(msg, UserWarning)
```



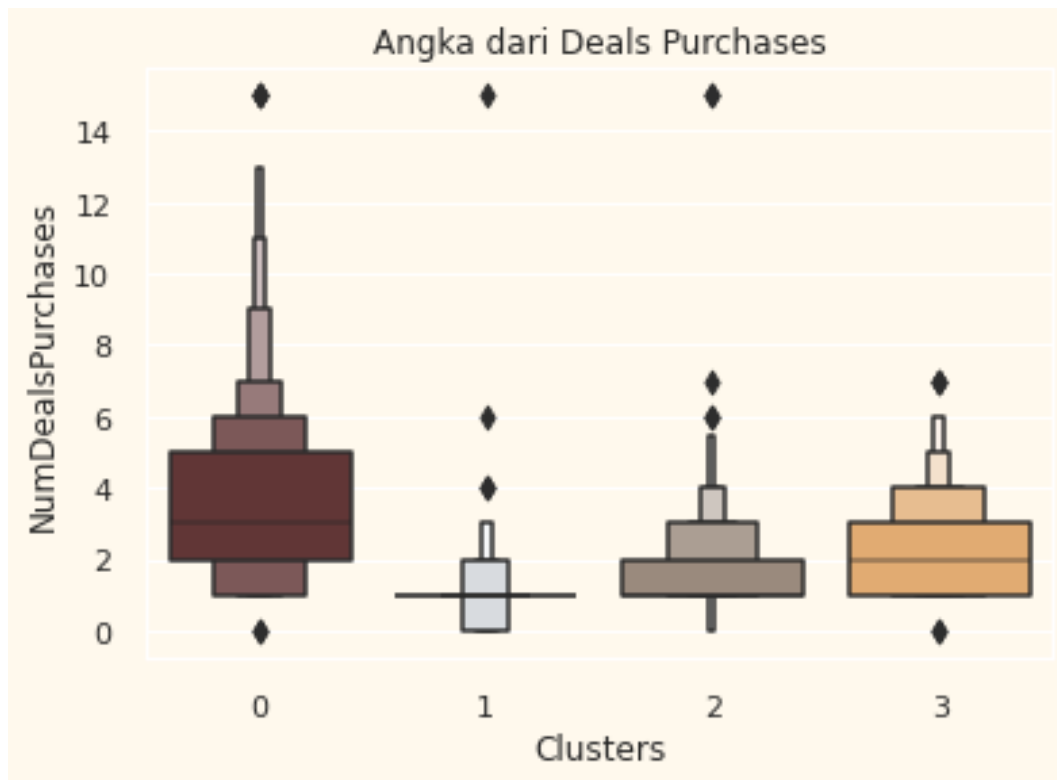
```
# Membuat fitur untuk mendapatkan sejumlah promosi yang
diterima/accepted promotions
data["Total_Promos"] = data["AcceptedCmp1"]+ data["AcceptedCmp2"]+
data["AcceptedCmp3"]+ data["AcceptedCmp4"]+ data["AcceptedCmp5"]

# plotting Jumlah dari total kampanye yang diterima.
plt.figure()
pl = sns.countplot(x=data["Total_Promos"],hue=data["Clusters"],
palette= pal)
pl.set_title("Jumlah dari Promotion Accepted")
pl.set_xlabel("Angka/number dari Total Accepted Promotions")
plt.show()
```





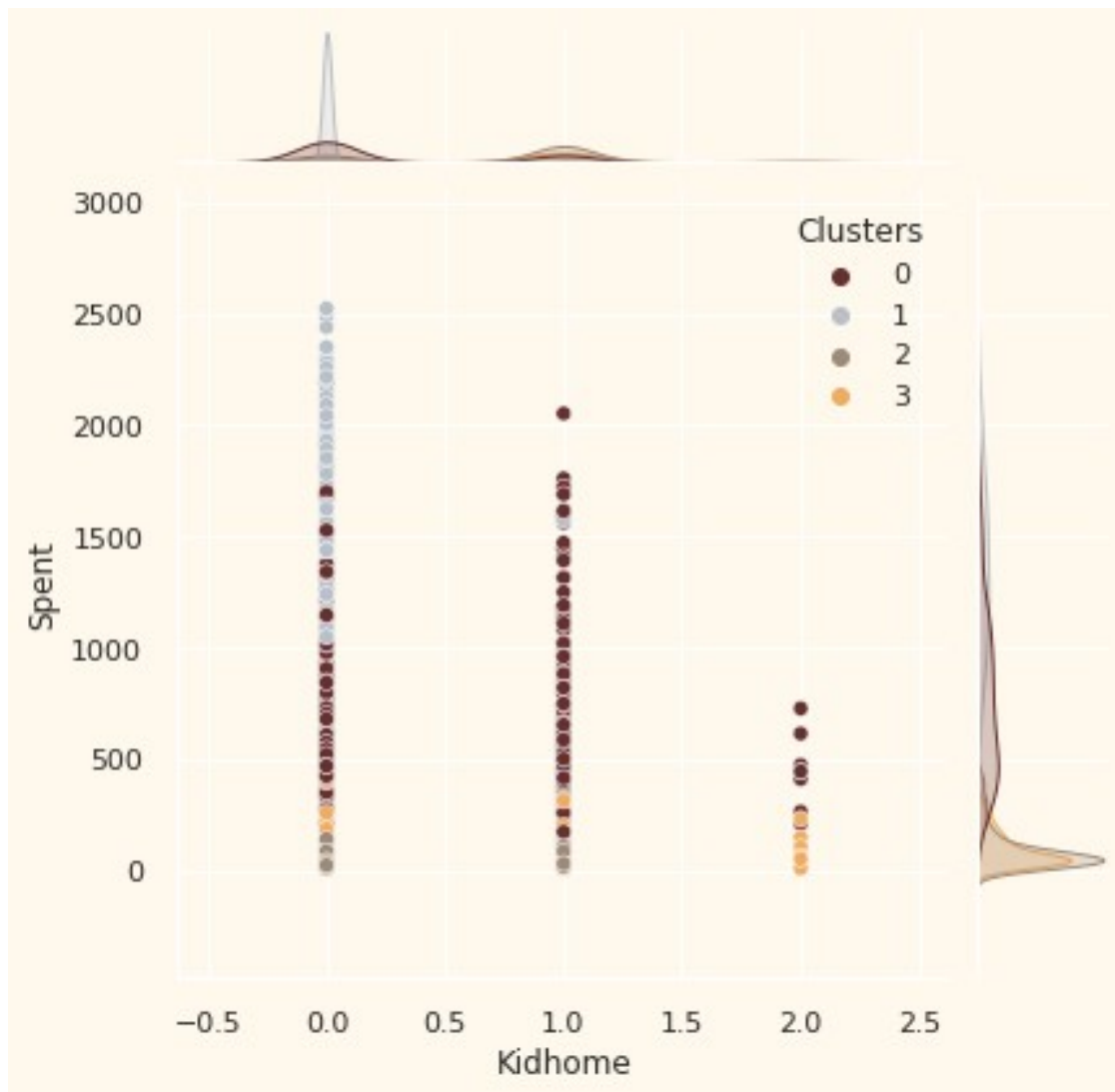
```
#Plotting jumlah transaksi yang dibeli / deals purchased
plt.figure()
pl=sns.boxenplot(y=data["NumDealsPurchases"],x=data["Clusters"],
palette= pal)
pl.set_title("Angka dari Deals Purchases")
plt.show()
```



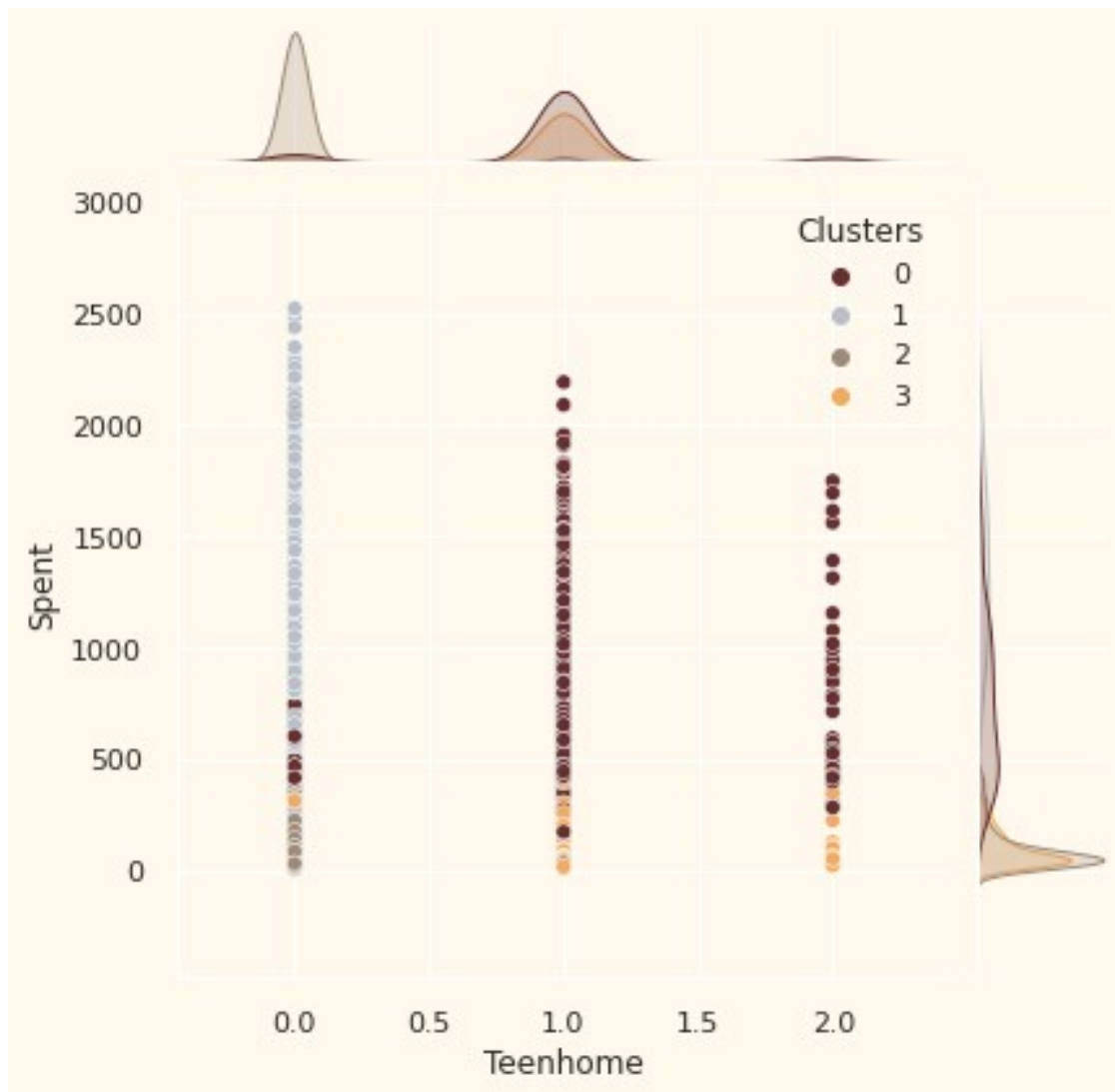
*# untuk detail lebih lanjut tentang gaya pembelian/purchasing style*  
 Places = [ "Kidhome", "Teenhome", "Customer\_For", "Age", "Children",  
 "Family\_Size", "Is\_Parent", "Education", "Living\_With"]

```
for i in Places:
    plt.figure()
    sns.jointplot(x=data[i], y = data["Spent"], hue=data["Clusters"],
    palette= pal)
    plt.show()
```

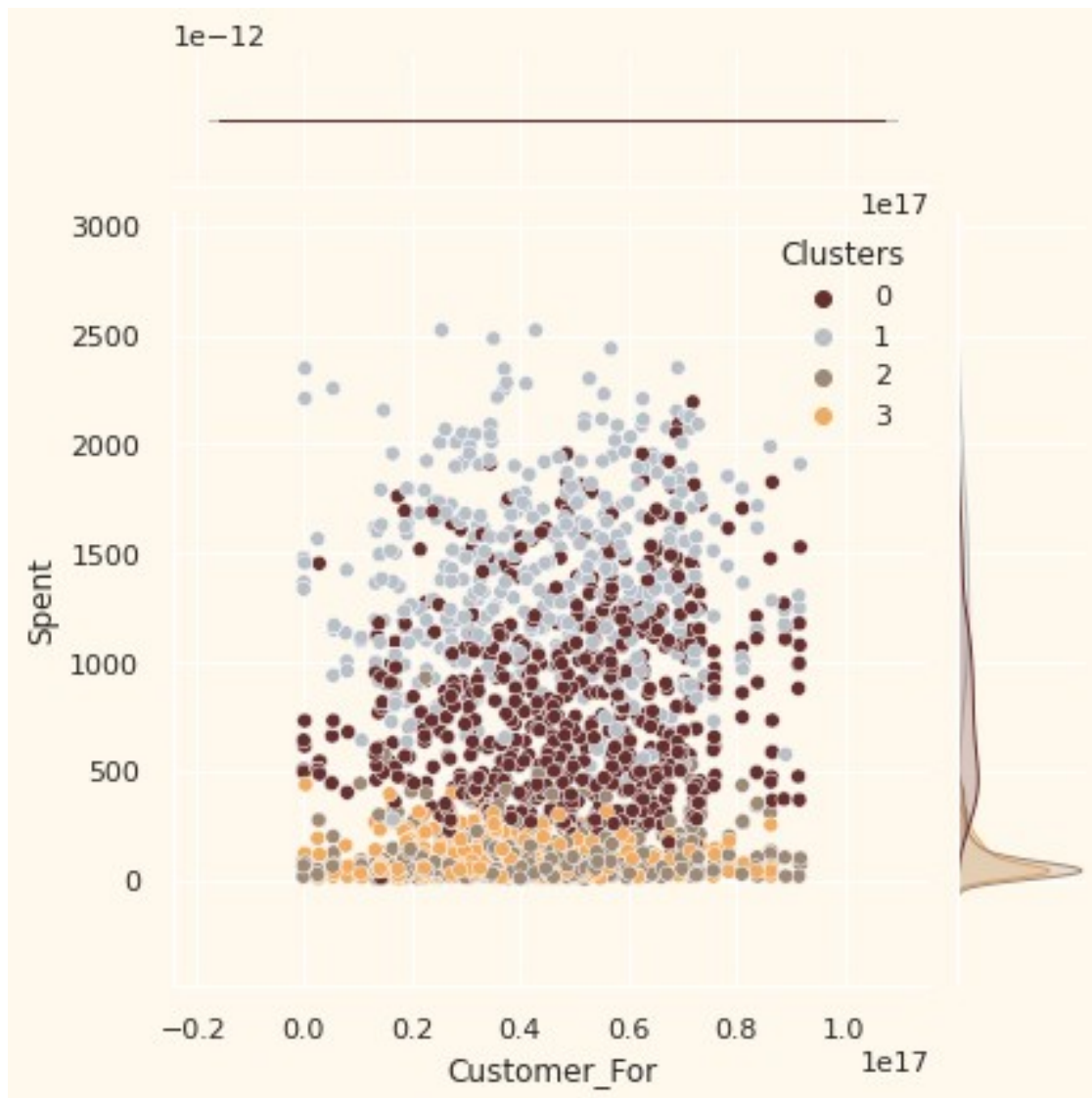
<Figure size 432x288 with 0 Axes>



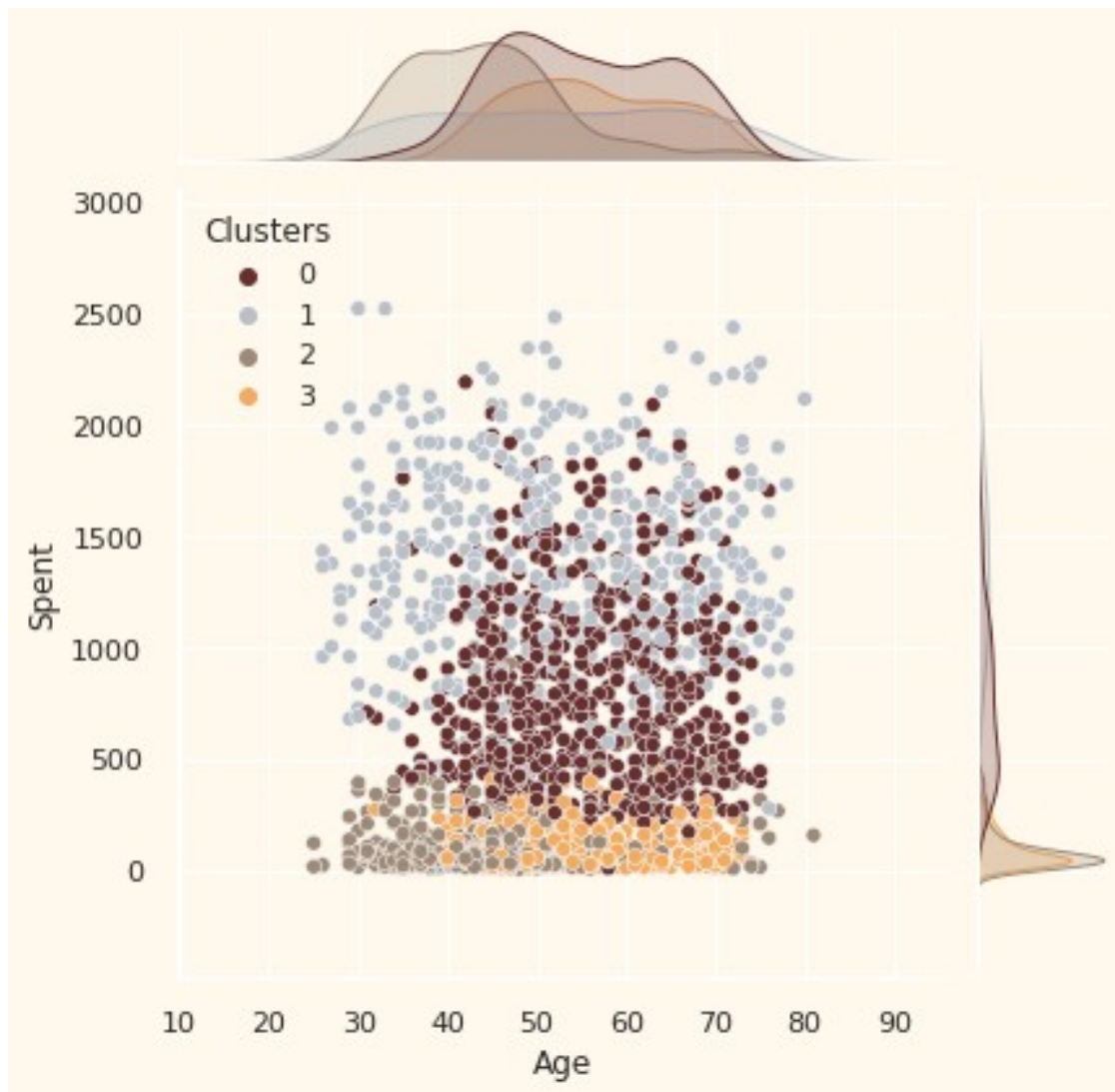
<Figure size 432x288 with 0 Axes>



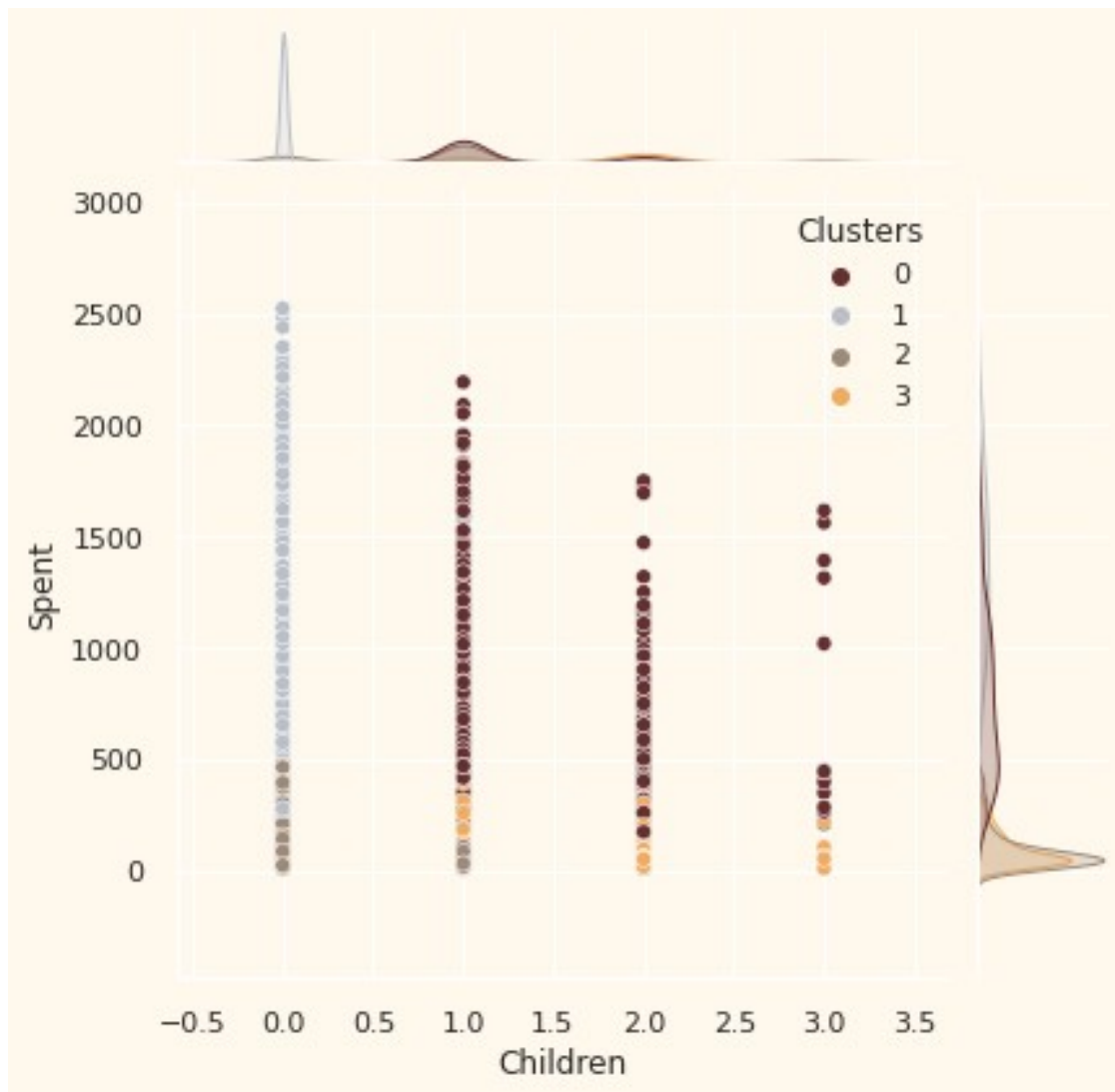
<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>

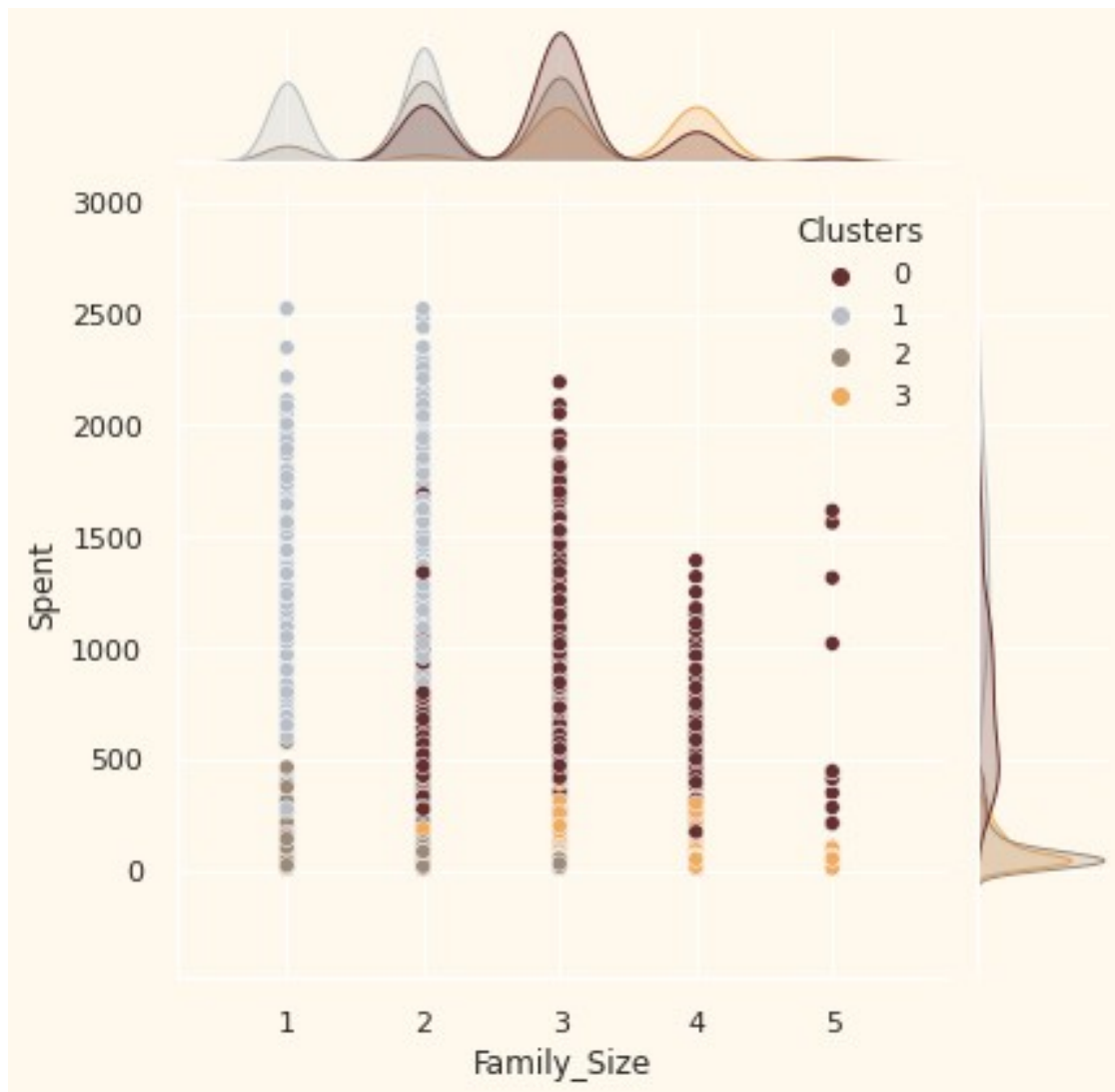


<Figure size 432x288 with 0 Axes>

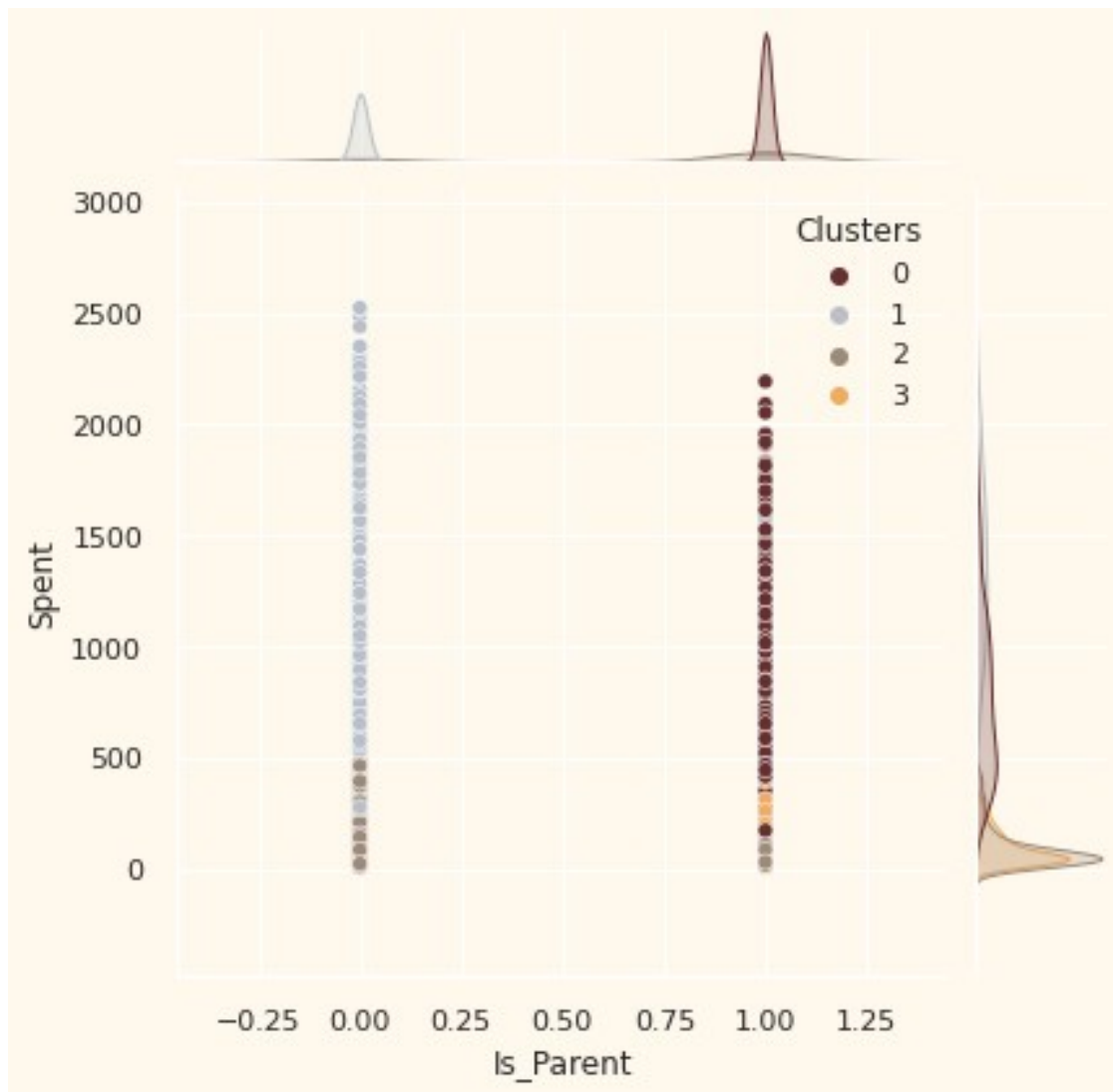


<Figure size 432x288 with 0 Axes>

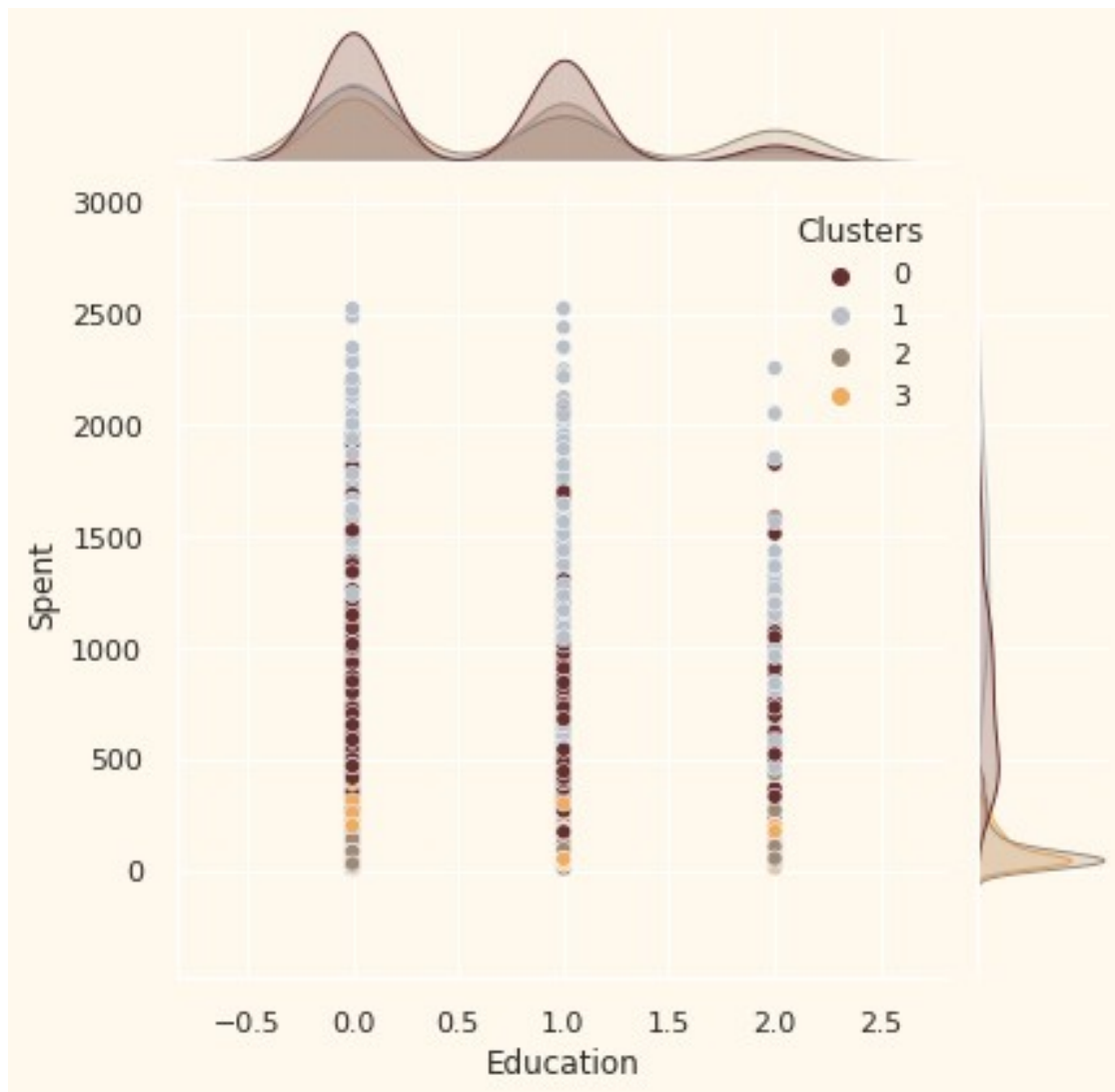




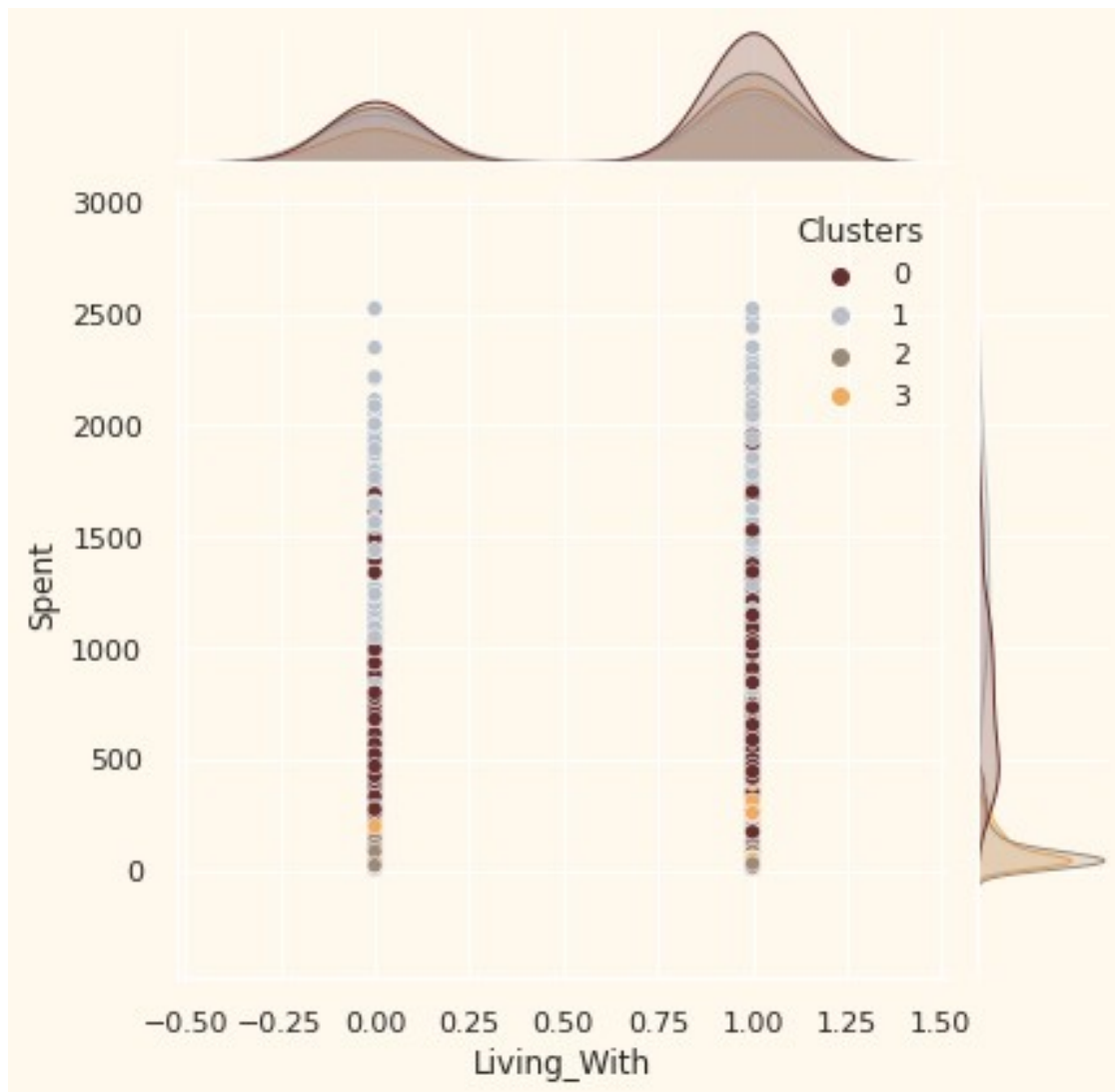
<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>

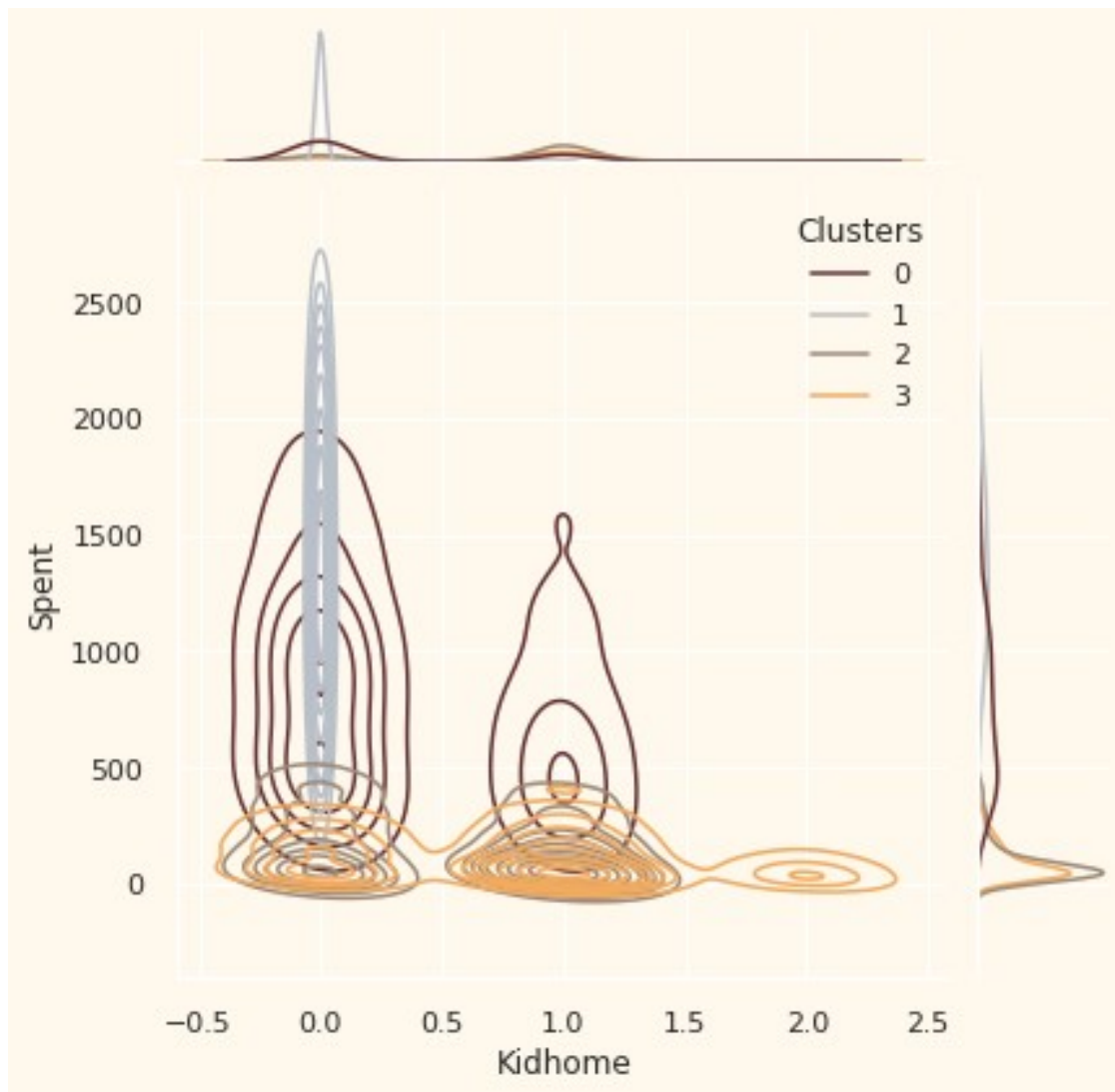


# Latihan 10

```
Personal = [ "Kidhome", "Teenhome", "Customer_For", "Age", "Children",
"Family_Size", "Is_Parent", "Education", "Living_With"]
```

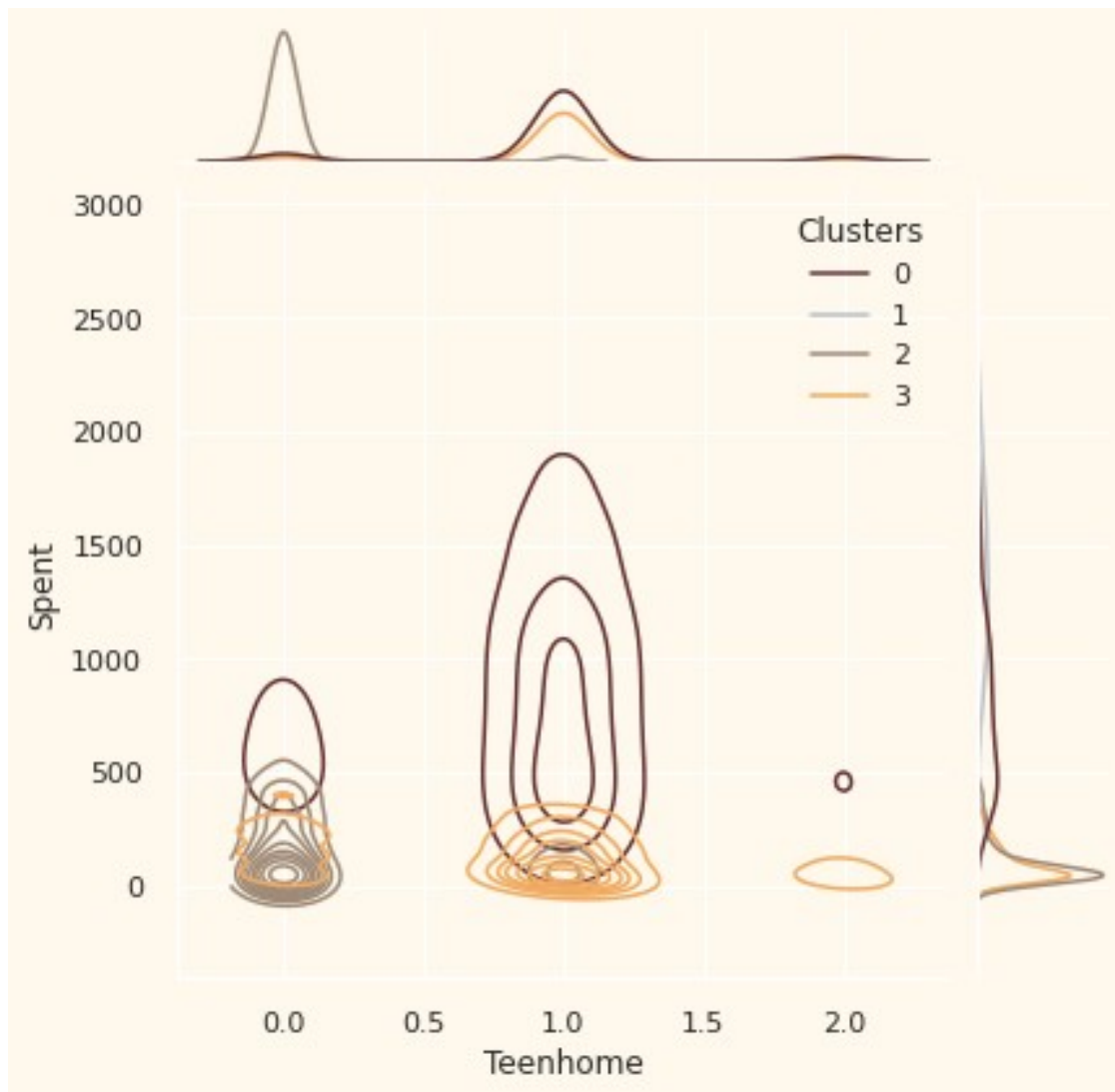
```
for i in Personal:
    plt.figure()
    sns.jointplot(x=data[i], y=data["Spent"], hue =data["Clusters"],
kind="kde", palette=pal)
    plt.show()
```

<Figure size 432x288 with 0 Axes>

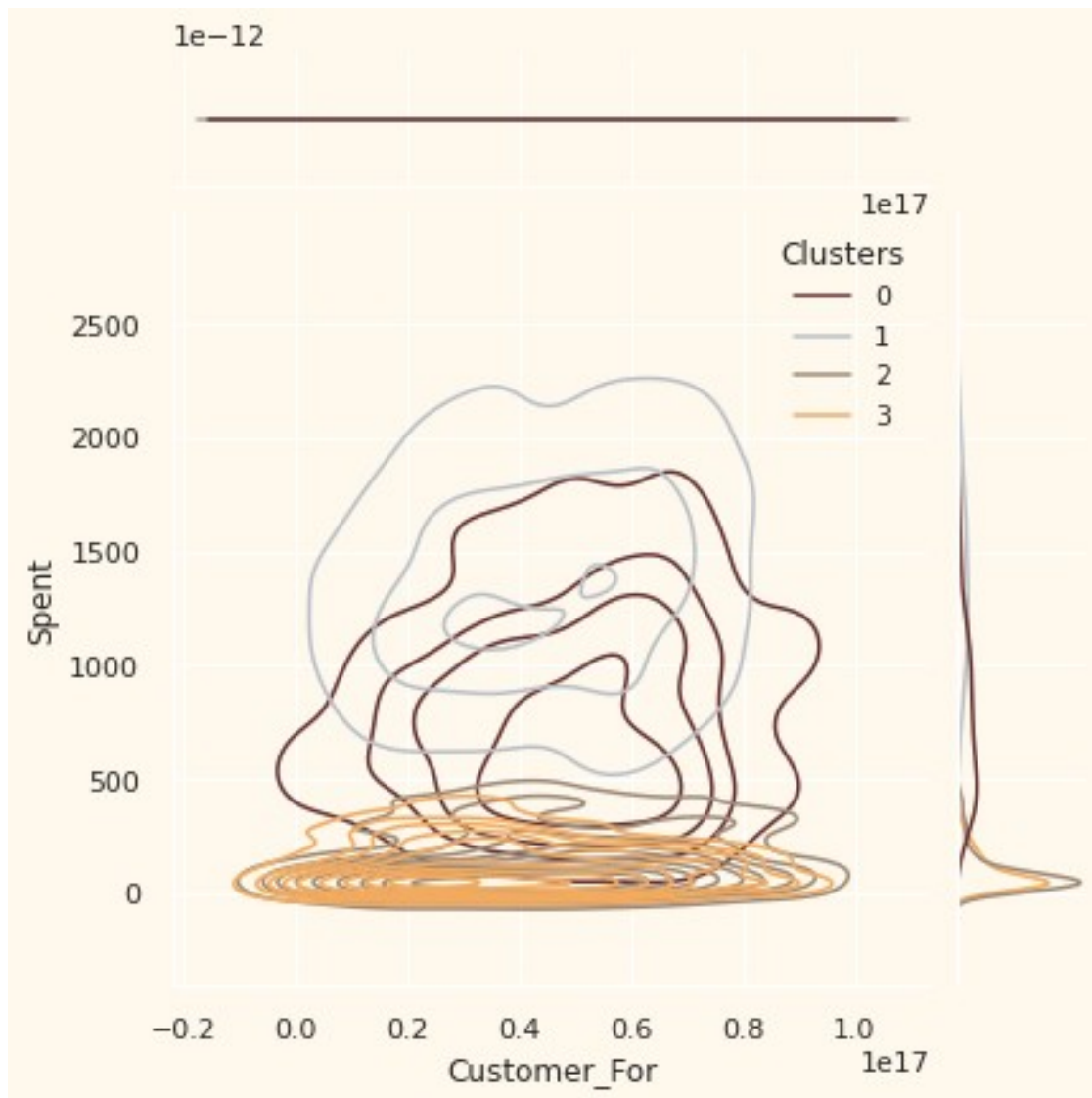


```
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:316:  
UserWarning: Dataset has 0 variance; skipping density estimate. Pass  
`warn_singular=False` to disable this warning.  
warnings.warn(msg, UserWarning)
```

<Figure size 432x288 with 0 Axes>

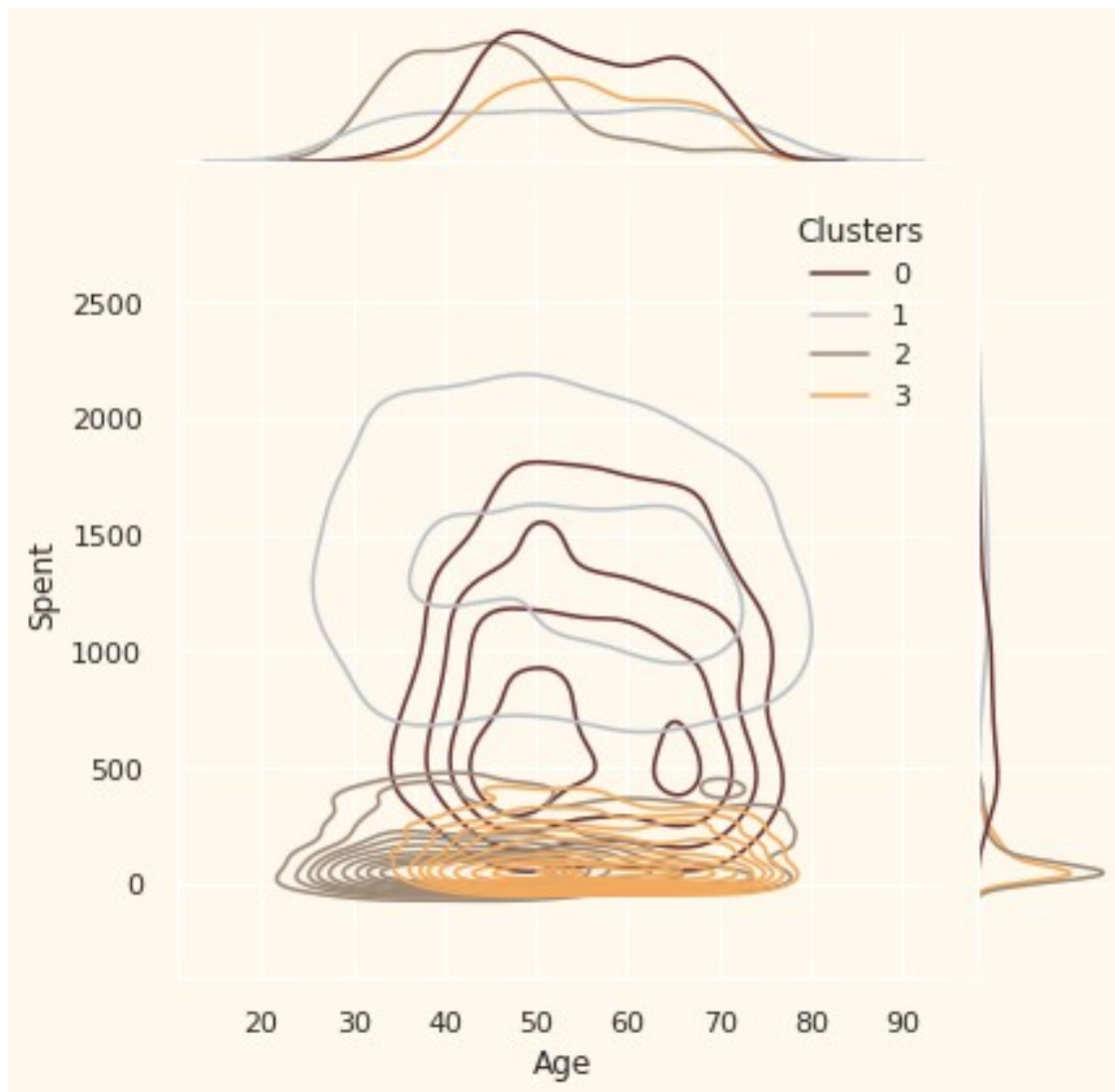


<Figure size 432x288 with 0 Axes>

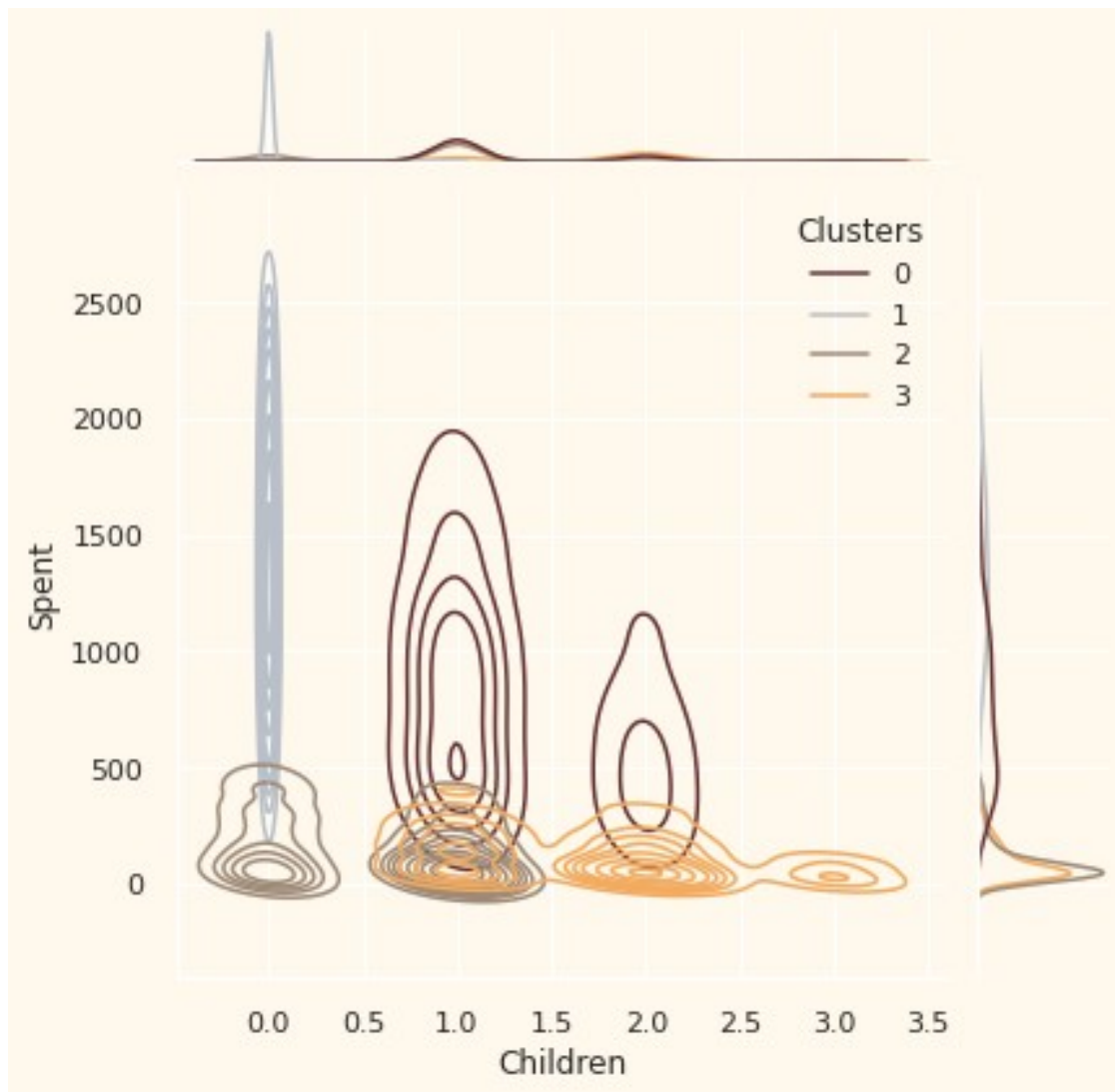


<Figure size 432x288 with 0 Axes>

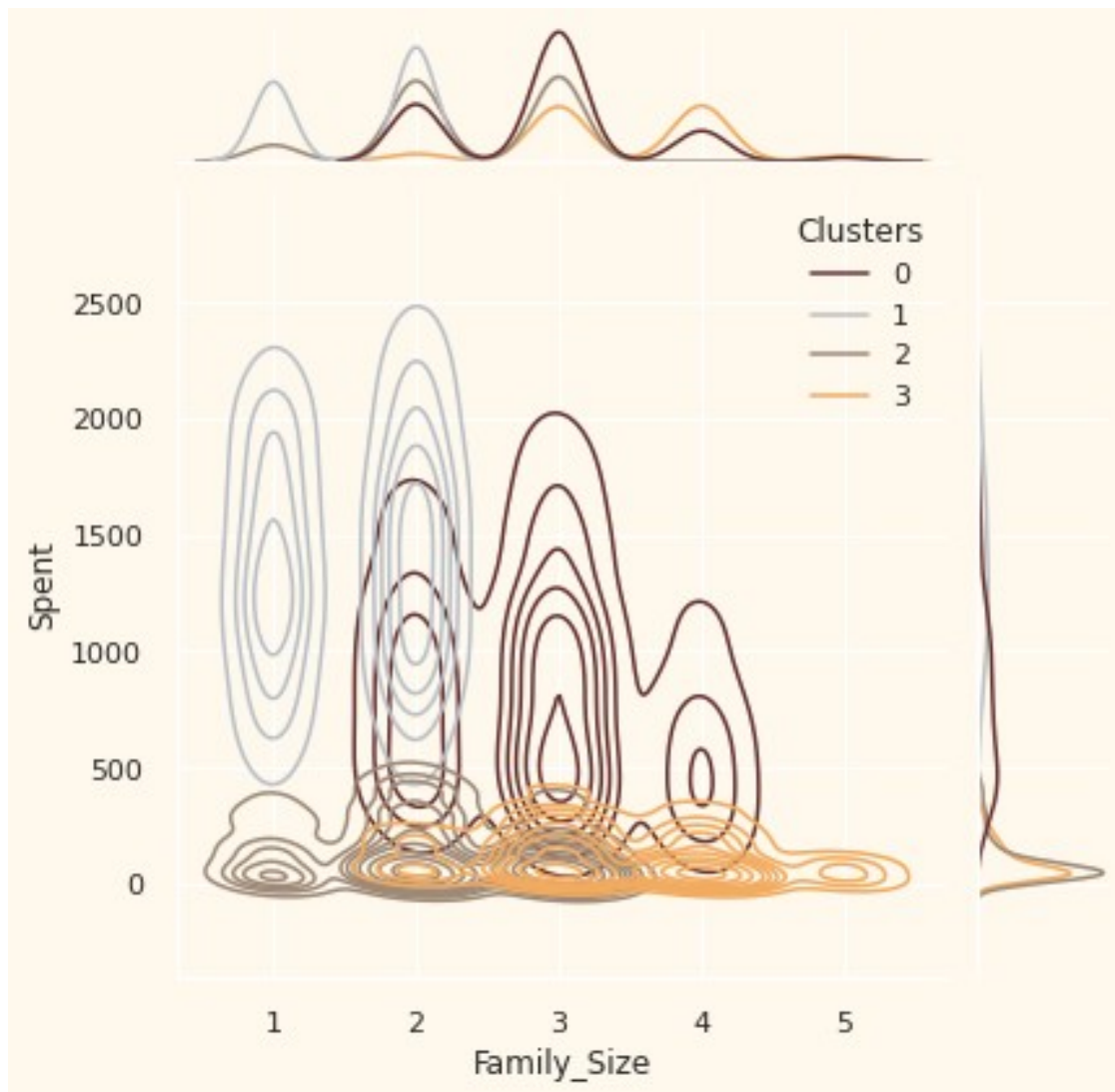




<Figure size 432x288 with 0 Axes>

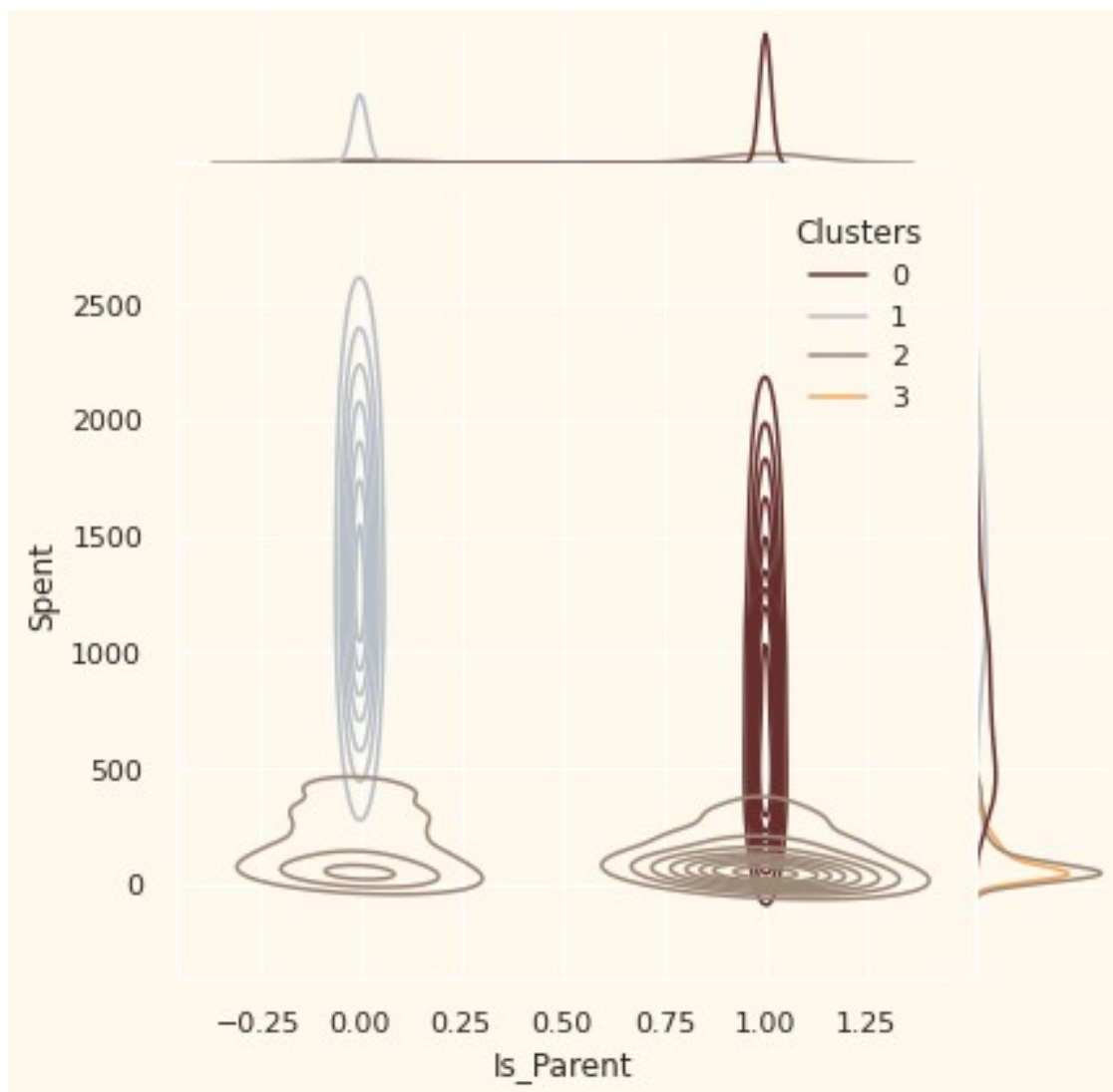


<Figure size 432x288 with 0 Axes>

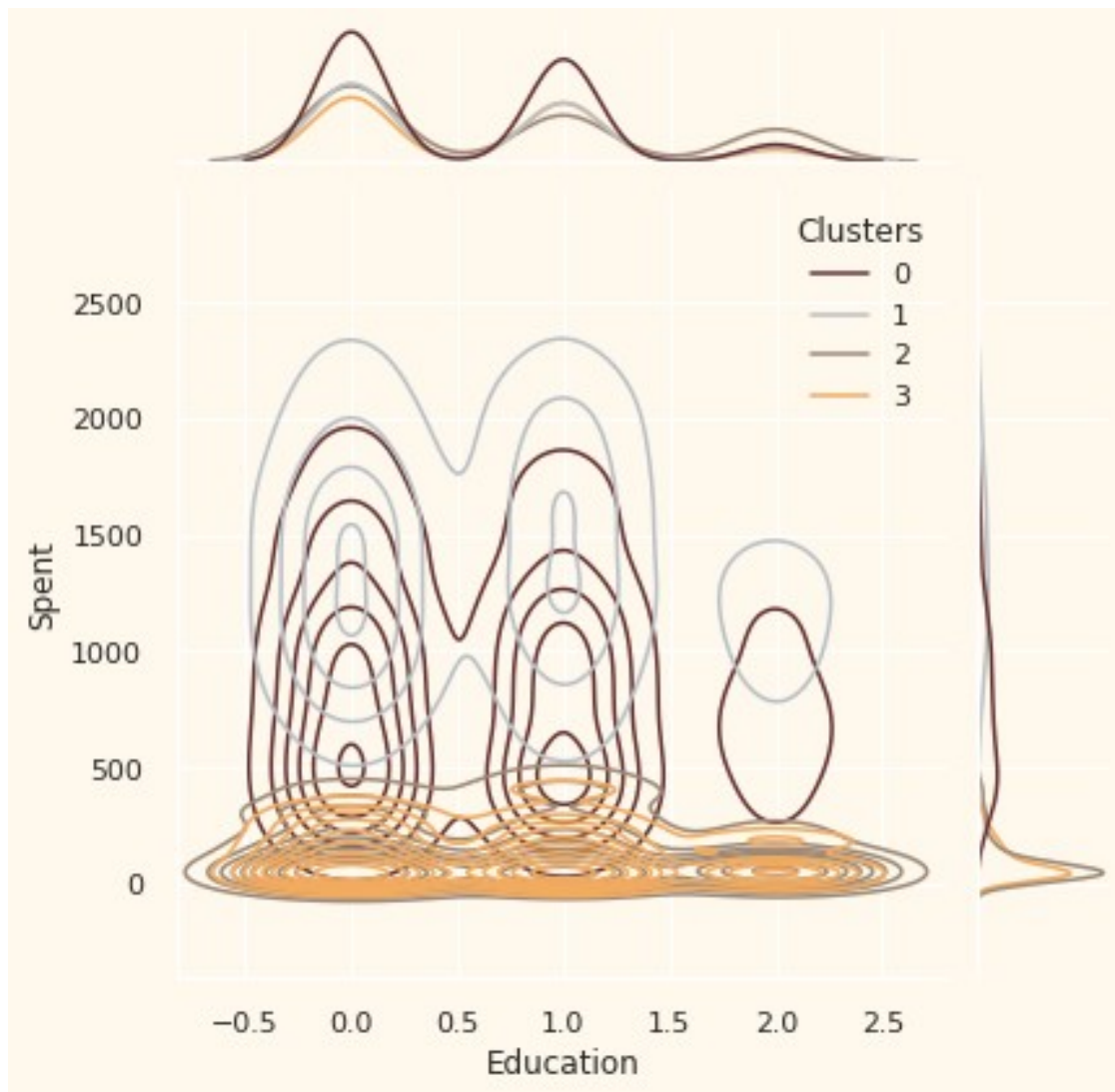


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