In [13]:
```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
pd.set_option('display.max_columns', None)
%matplotlib inline
sns.set_context('notebook')
sns.set_style('whitegrid')
sns.set_palette('Blues_r')

# turn off warnings for final notebook
import warnings
warnings.filterwarnings('ignore')
```

In [14]:
```
df=pd.read_csv("ifood_df.csv")
```

In [15]:
```
df
```

Out[15]:

	Income	Kidhome	Teenhome	Recency	MntWines	MntFruits	MntMeatProducts	MntFishProducts	MntSweetProducts	MntGoldPro
0	58138.0	0	0	58	635	88	546	172	88	
1	46344.0	1	1	38	11	1	6	2	1	
2	71613.0	0	0	26	426	49	127	111	21	
3	26646.0	1	0	26	11	4	20	10	3	
4	58293.0	1	0	94	173	43	118	46	27	
...	...	...	...	...	...	...	...	...	...	
2200	61223.0	0	1	46	709	43	182	42	118	2
2201	64014.0	2	1	56	406	0	30	0	0	
2202	56981.0	0	0	91	908	48	217	32	12	
2203	69245.0	0	1	8	428	30	214	80	30	
2204	52869.0	1	1	40	84	3	61	2	1	

2205 rows × 39 columns

In [16]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2205 entries, 0 to 2204
Data columns (total 39 columns):
 #   Column                Non-Null Count  Dtype
---  ------                --------------  -----
 0   Income                2205 non-null   float64
 1   Kidhome               2205 non-null   int64
 2   Teenhome              2205 non-null   int64
 3   Recency               2205 non-null   int64
 4   MntWines              2205 non-null   int64
 5   MntFruits             2205 non-null   int64
 6   MntMeatProducts       2205 non-null   int64
 7   MntFishProducts       2205 non-null   int64
 8   MntSweetProducts      2205 non-null   int64
 9   MntGoldProds          2205 non-null   int64
 10  NumDealsPurchases     2205 non-null   int64
 11  NumWebPurchases       2205 non-null   int64
 12  NumCatalogPurchases   2205 non-null   int64
 13  NumStorePurchases     2205 non-null   int64
 14  NumWebVisitsMonth     2205 non-null   int64
 15  AcceptedCmp3          2205 non-null   int64
 16  AcceptedCmp4          2205 non-null   int64
 17  AcceptedCmp5          2205 non-null   int64
 18  AcceptedCmp1          2205 non-null   int64
 19  AcceptedCmp2          2205 non-null   int64
 20  Complain              2205 non-null   int64
 21  Z_CostContact         2205 non-null   int64
 22  Z_Revenue             2205 non-null   int64
 23  Response              2205 non-null   int64
 24  Age                   2205 non-null   int64
 25  Customer_Days         2205 non-null   int64
 26  marital_Divorced      2205 non-null   int64
 27  marital_Married       2205 non-null   int64
 28  marital_Single        2205 non-null   int64
 29  marital_Together      2205 non-null   int64
 30  marital_Widow         2205 non-null   int64
 31  education_2n Cycle    2205 non-null   int64
 32  education_Basic       2205 non-null   int64
 33  education_Graduation  2205 non-null   int64
 34  education_Master      2205 non-null   int64
 35  education_PhD         2205 non-null   int64
```

```
 36  MntTotal              2205 non-null   int64
 37  MntRegularProds       2205 non-null   int64
 38  AcceptedCmpOverall    2205 non-null   int64
dtypes: float64(1), int64(38)
memory usage: 672.0 KB
```

In [17]:
```
df.describe()
```

Out[17]:

	Income	Kidhome	Teenhome	Recency	MntWines	MntFruits	MntMeatProducts	MntFishProducts	MntSweetPro
count	2205.000000	2205.000000	2205.000000	2205.000000	2205.000000	2205.000000	2205.000000	2205.000000	2205.0
mean	51622.094785	0.442177	0.506576	49.009070	306.164626	26.403175	165.312018	37.756463	27.1
std	20713.063826	0.537132	0.544380	28.932111	337.493839	39.784484	217.784507	54.824635	41.1
min	1730.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0
25%	35196.000000	0.000000	0.000000	24.000000	24.000000	2.000000	16.000000	3.000000	1.0
50%	51287.000000	0.000000	0.000000	49.000000	178.000000	8.000000	68.000000	12.000000	8.0
75%	68281.000000	1.000000	1.000000	74.000000	507.000000	33.000000	232.000000	50.000000	34.0
max	113734.000000	2.000000	2.000000	99.000000	1493.000000	199.000000	1725.000000	259.000000	262.0

In [18]:
```
df.shape
```

Out[18]:
```
(2205, 39)
```

In [19]:
```
df.transpose
```

Out[19]: <bound method DataFrame.transpose of   Income Kidhome Teenhome Recency MntWines MntFruits \
0   58138.0   0   0   58   635   88
1   46344.0   1   1   38   11   1
2   71613.0   0   0   26   426   49
3   26646.0   1   0   26   11   4
4   58293.0   1   0   94   173   43
...   .
2200 61223.0   0   1   46   709   43
2201 64014.0   2   1   56   406   0
2202 56981.0   0   0   91   908   48
2203 69245.0   0   1   8   428   30
2204 52869.0   1   1   40   84   3

      MntMeatProducts MntFishProducts MntSweetProducts MntGoldProds \
0   546   172   88   88
1   6   2   1   6
2   127   111   21   42
3   20   10   3   5
4   118   46   27   15
...   .
2200   182   42   118   247
2201   30   0   0   8
2202   217   32   12   24
2203   214   80   30   61
2204   61   2   1   21

      NumDealsPurchases NumWebPurchases NumCatalogPurchases \
0   3   8   10
1   2   1   1
2   1   8   2
3   2   2   0
4   5   5   3
...   .
2200   2   9   3
2201   7   8   2
2202   1   2   3
2203   2   6   5
2204   3   3   1

      NumStorePurchases NumWebVisitsMonth AcceptedCmp3 AcceptedCmp4 \

```
0                    4                7            0            0
1                    2                5            0            0
2                   10                4            0            0
3                    4                6            0            0
4                    6                5            0            0
...                ...              ...          ...          ...
2200                 4                5            0            0
2201                 5                7            0            0
2202                13                6            0            1
2203                10                3            0            0
2204                 4                7            0            0

      AcceptedCmp5  AcceptedCmp1  AcceptedCmp2  Complain  Z_CostContact  \
0                0             0             0         0              3
1                0             0             0         0              3
2                0             0             0         0              3
3                0             0             0         0              3
4                0             0             0         0              3
...            ...           ...           ...       ...            ...
2200             0             0             0         0              3
2201             0             1             0         0              3
2202             0             0             0         0              3
2203             0             0             0         0              3
2204             0             0             0         0              3

      Z_Revenue  Response  Age  Customer_Days  marital_Divorced  \
0            11         1   63           2822                 0
1            11         0   66           2272                 0
2            11         0   55           2471                 0
3            11         0   36           2298                 0
4            11         0   39           2320                 0
...         ...       ...  ...            ...               ...
2200         11         0   53           2540                 0
2201         11         0   74           2178                 0
2202         11         0   39           2314                 1
2203         11         0   64           2315                 0
2204         11         1   66           2781                 0

      marital_Married  marital_Single  marital_Together  marital_Widow  \
0                   0               1                 0              0
1                   0               1                 0              0
```

```
2                   0              0                 1              0
3                   0              0                 1              0
4                   1              0                 0              0
...               ...            ...               ...            ...
2200                1              0                 0              0
2201                0              0                 1              0
2202                0              0                 0              0
2203                0              0                 1              0
2204                1              0                 0              0

      education_2n Cycle  education_Basic  education_Graduation  \
0                      0                0                     1
1                      0                0                     1
2                      0                0                     1
3                      0                0                     1
4                      0                0                     0
...                  ...              ...                   ...
2200                   0                0                     1
2201                   0                0                     0
2202                   0                0                     1
2203                   0                0                     0
2204                   0                0                     0

      education_Master  education_PhD  MntTotal  MntRegularProds  \
0                    0              0      1529             1441
1                    0              0        21               15
2                    0              0       734              692
3                    0              0        48               43
4                    0              1       407              392
...                ...            ...       ...              ...
2200                 0              0      1094              847
2201                 0              1       436              428
2202                 0              0      1217             1193
2203                 1              0       782              721
2204                 0              1       151              130

      AcceptedCmpOverall
0                      0
1                      0
2                      0
3                      0
```

```
4                     0
...                 ...
2200                  0
2201                  1
2202                  1
2203                  0
2204                  0

[2205 rows x 39 columns]>
```

In [10]:
```
df.head()
```

Out[10]:

	Income	Kidhome	Teenhome	Recency	MntWines	MntFruits	MntMeatProducts	MntFishProducts	MntSweetProducts	MntGoldProds
0	58138.0	0	0	58	635	88	546	172	88	88
1	46344.0	1	1	38	11	1	6	2	1	6
2	71613.0	0	0	26	426	49	127	111	21	42
3	26646.0	1	0	26	11	4	20	10	3	5
4	58293.0	1	0	94	173	43	118	46	27	15

5 rows × 39 columns

In [12]:
```
df.shape[0]
```

Out[12]:
```
2205
```

In [20]:
```
df.isnull().sum().sort_values(ascending=True)
```

Out[20]:
```
Income                0
Z_CostContact         0
Z_Revenue             0
Response              0
Age                   0
Customer_Days         0
marital_Divorced      0
marital_Married       0
Complain              0
marital_Single        0
marital_Widow         0
education_2n Cycle    0
education_Basic       0
education_Graduation  0
education_Master      0
education_PhD         0
MntTotal              0
marital_Together      0
MntRegularProds       0
AcceptedCmp2          0
AcceptedCmp5          0
Kidhome               0
Teenhome              0
Recency               0
MntWines              0
MntFruits             0
MntMeatProducts       0
MntFishProducts       0
AcceptedCmp1          0
MntSweetProducts      0
NumDealsPurchases     0
NumWebPurchases       0
NumCatalogPurchases   0
NumStorePurchases     0
NumWebVisitsMonth     0
AcceptedCmp3          0
AcceptedCmp4          0
MntGoldProds          0
AcceptedCmpOverall    0
dtype: int64
```

In [21]:
```
df.isnull().sum().sort_values(ascending=False)
```

Out[21]:
```
Income                 0
marital_Together       0
Z_Revenue              0
Response               0
Age                    0
Customer_Days          0
marital_Divorced       0
marital_Married        0
marital_Single         0
marital_Widow          0
Complain               0
education_2n Cycle     0
education_Basic        0
education_Graduation   0
education_Master       0
education_PhD          0
MntTotal               0
MntRegularProds        0
Z_CostContact          0
AcceptedCmp2           0
Kidhome                0
MntGoldProds           0
Teenhome               0
Recency                0
MntWines               0
MntFruits              0
MntMeatProducts        0
MntFishProducts        0
MntSweetProducts       0
NumDealsPurchases      0
AcceptedCmp1           0
NumWebPurchases        0
NumCatalogPurchases    0
NumStorePurchases      0
NumWebVisitsMonth      0
AcceptedCmp3           0
AcceptedCmp4           0
AcceptedCmp5           0
AcceptedCmpOverall     0
dtype: int64
```

In [24]:
```
df = df.dropna()
```

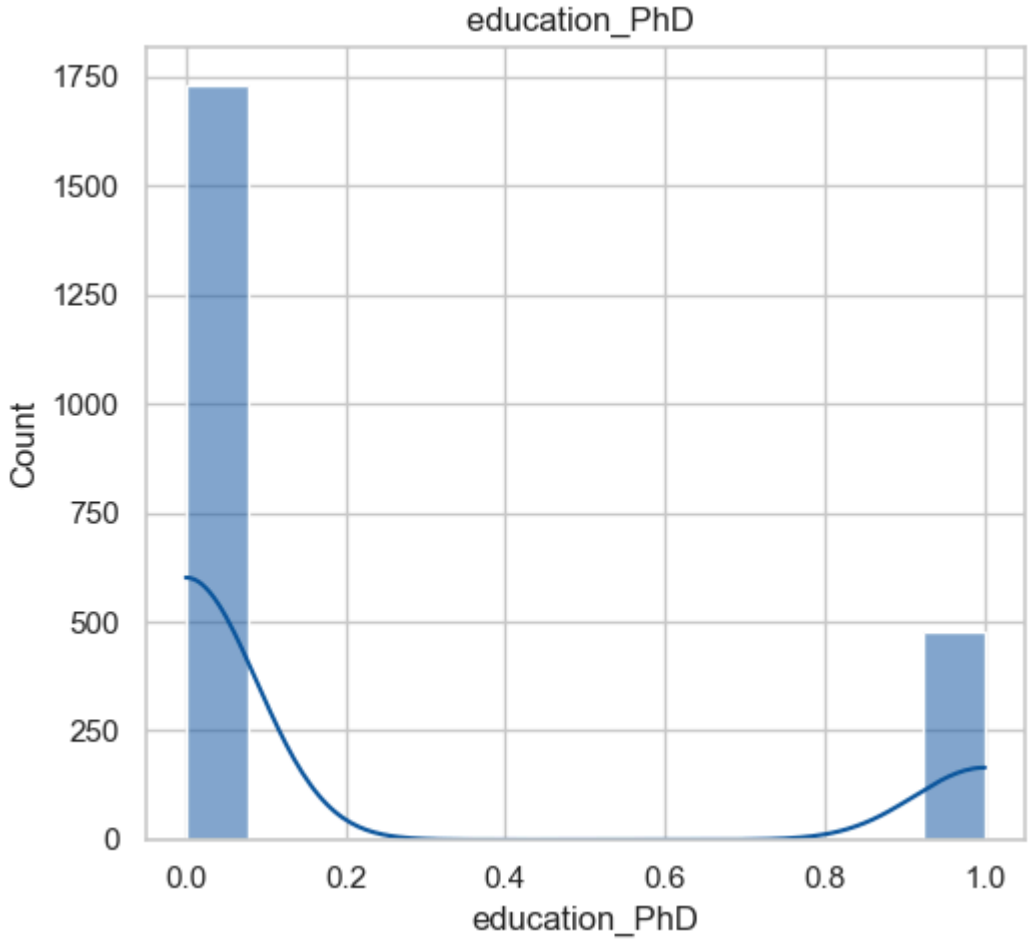
In [72]:
```
print(df_cleaned.describe())
```

```
           Kidhome     Teenhome      Recency     MntWines    MntFruits  \
count  2205.000000  2205.000000  2205.000000  2205.000000  2205.000000
mean      0.442177     0.506576    49.009070   306.164626    26.403175
std       0.537132     0.544380    28.932111   337.493839    39.784484
min       0.000000     0.000000     0.000000     0.000000     0.000000
25%       0.000000     0.000000    24.000000    24.000000     2.000000
50%       0.000000     0.000000    49.000000   178.000000     8.000000
75%       1.000000     1.000000    74.000000   507.000000    33.000000
max       2.000000     2.000000    99.000000  1493.000000   199.000000

       MntMeatProducts  MntFishProducts  MntSweetProducts  MntGoldProds  \
count      2205.000000      2205.000000       2205.000000   2205.000000
mean        165.312018        37.756463         27.128345     44.057143
std         217.784507        54.824635         41.130468     51.736211
min           0.000000         0.000000          0.000000      0.000000
25%          16.000000         3.000000          1.000000      9.000000
50%          68.000000        12.000000          8.000000     25.000000
75%         232.000000        50.000000         34.000000     56.000000
max        1725.000000       259.000000        262.000000    321.000000

       NumDealsPurchases  NumWebPurchases  NumCatalogPurchases  \
count        2205.000000      2205.000000          2205.000000
mean            2.318367         4.100680             2.645351
std             1.886107         2.737424             2.798647
min             0.000000         0.000000             0.000000
25%             1.000000         2.000000             0.000000
50%             2.000000         4.000000             2.000000
75%             3.000000         6.000000             4.000000
max            15.000000        27.000000            28.000000

       NumStorePurchases  NumWebVisitsMonth  AcceptedCmp3  AcceptedCmp4  \
count        2205.000000        2205.000000   2205.000000   2205.000000
mean            5.823583           5.336961      0.073923      0.074376
std             3.241796           2.413535      0.261705      0.262442
min             0.000000           0.000000      0.000000      0.000000
25%             3.000000           3.000000      0.000000      0.000000
50%             5.000000           6.000000      0.000000      0.000000
75%             8.000000           7.000000      0.000000      0.000000
max            13.000000          20.000000      1.000000      1.000000

       AcceptedCmp5  AcceptedCmp1  AcceptedCmp2     Complain  Z_CostContact  \
```

```
count   2205.000000   2205.000000   2205.000000  2205.000000         2205.0
mean       0.073016      0.064399      0.013605     0.009070            3.0
std        0.260222      0.245518      0.115872     0.094827            0.0
min        0.000000      0.000000      0.000000     0.000000            3.0
25%        0.000000      0.000000      0.000000     0.000000            3.0
50%        0.000000      0.000000      0.000000     0.000000            3.0
75%        0.000000      0.000000      0.000000     0.000000            3.0
max        1.000000      1.000000      1.000000     1.000000            3.0

       Z_Revenue    Response          Age  Customer_Days  marital_Divorced  \
count     2205.0  2205.00000  2205.000000    2205.000000       2205.000000
mean        11.0     0.15102    51.095692    2512.718367          0.104308
std          0.0     0.35815    11.705801     202.563647          0.305730
min         11.0     0.00000    24.000000    2159.000000          0.000000
25%         11.0     0.00000    43.000000    2339.000000          0.000000
50%         11.0     0.00000    50.000000    2515.000000          0.000000
75%         11.0     0.00000    61.000000    2688.000000          0.000000
max         11.0     1.00000    80.000000    2858.000000          1.000000

       marital_Married  marital_Single  marital_Together  marital_Widow  \
count      2205.000000     2205.000000       2205.000000    2205.000000
mean          0.387302        0.216327          0.257596       0.034467
std           0.487244        0.411833          0.437410       0.182467
min           0.000000        0.000000          0.000000       0.000000
25%           0.000000        0.000000          0.000000       0.000000
50%           0.000000        0.000000          0.000000       0.000000
75%           1.000000        0.000000          1.000000       0.000000
max           1.000000        1.000000          1.000000       1.000000

       education_2n Cycle  education_Basic  education_Graduation  \
count         2205.000000      2205.000000           2205.000000
mean             0.089796         0.024490              0.504762
std              0.285954         0.154599              0.500091
min              0.000000         0.000000              0.000000
25%              0.000000         0.000000              0.000000
50%              0.000000         0.000000              1.000000
75%              0.000000         0.000000              1.000000
max              1.000000         1.000000              1.000000

       education_Master  education_PhD     MntTotal  MntRegularProds  \
count       2205.000000    2205.000000  2205.000000      2205.000000
```

```
mean           0.165079       0.215873   562.764626       518.707483
std            0.371336       0.411520   575.936911       553.847248
min            0.000000       0.000000     4.000000      -283.000000
25%            0.000000       0.000000    56.000000        42.000000
50%            0.000000       0.000000   343.000000       288.000000
75%            0.000000       0.000000   964.000000       884.000000
max            1.000000       1.000000  2491.000000      2458.000000

       AcceptedCmpOverall      Cluster
count          2205.00000  2205.000000
mean              0.29932     1.494331
std               0.68044     1.073349
min               0.00000     0.000000
25%               0.00000     1.000000
50%               0.00000     1.000000
75%               0.00000     2.000000
max               4.00000     3.000000
```

In [77]:
```
plt.figure(figsize=(10, 5))

plt.subplot(1, 2, 1)
sns.histplot(df_cleaned['education_PhD'], kde=True)
plt.title('education_PhD ')

plt.tight_layout()
plt.show()
```



In [27]:
```
df_cleaned = df.drop(columns=['Income'], axis=1)

# View cleaned dataset
print(df_cleaned.head())
```

	Kidhome	Teenhome	Recency	MntWines	MntFruits	MntMeatProducts	\
0	0	0	58	635	88	546	
1	1	1	38	11	1	6	
2	0	0	26	426	49	127	
3	1	0	26	11	4	20	
4	1	0	94	173	43	118	

	MntFishProducts	MntSweetProducts	MntGoldProds	NumDealsPurchases	\
0	172	88	88	3	
1	2	1	6	2	
2	111	21	42	1	
3	10	3	5	2	
4	46	27	15	5	

	NumWebPurchases	NumCatalogPurchases	NumStorePurchases	NumWebVisitsMonth	\
0	8	10	4	7	
1	1	1	2	5	
2	8	2	10	4	
3	2	0	4	6	
4	5	3	6	5	

	AcceptedCmp3	AcceptedCmp4	AcceptedCmp5	AcceptedCmp1	AcceptedCmp2	\
0	0	0	0	0	0	
1	0	0	0	0	0	
2	0	0	0	0	0	
3	0	0	0	0	0	
4	0	0	0	0	0	

	Complain	Z_CostContact	Z_Revenue	Response	Age	Customer_Days	\
0	0	3	11	1	63	2822	
1	0	3	11	0	66	2272	
2	0	3	11	0	55	2471	
3	0	3	11	0	36	2298	
4	0	3	11	0	39	2320	

	marital_Divorced	marital_Married	marital_Single	marital_Together	\
0	0	0	1	0	
1	0	0	1	0	
2	0	0	0	1	
3	0	0	0	1	
4	0	1	0	0	

```
   marital_Widow  education_2n Cycle  education_Basic  education_Graduation  \
0              0                   0                0                     1
1              0                   0                0                     1
2              0                   0                0                     1
3              0                   0                0                     1
4              0                   0                0                     0

   education_Master  education_PhD  MntTotal  MntRegularProds  \
0                 0              0      1529             1441
1                 0              0        21               15
2                 0              0       734              692
3                 0              0        48               43
4                 0              1       407              392

   AcceptedCmpOverall
0                   0
1                   0
2                   0
3                   0
4                   0
```

In [28]:

```python
print(df_cleaned.describe())

# Visualizing distribution of annual income and spending score
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
sns.histplot(df_cleaned['Income'], kde=True)
plt.title('Annual Income Distribution')

plt.subplot(1, 2, 2)
sns.histplot(df_cleaned['NumStorePurchases'], kde=True)
plt.title('Spending Score Distribution')

plt.tight_layout()
plt.show()
```

```
           Kidhome     Teenhome      Recency     MntWines    MntFruits  \
count  2205.000000  2205.000000  2205.000000  2205.000000  2205.000000
mean      0.442177     0.506576    49.009070   306.164626    26.403175
std       0.537132     0.544380    28.932111   337.493839    39.784484
min       0.000000     0.000000     0.000000     0.000000     0.000000
25%       0.000000     0.000000    24.000000    24.000000     2.000000
50%       0.000000     0.000000    49.000000   178.000000     8.000000
75%       1.000000     1.000000    74.000000   507.000000    33.000000
max       2.000000     2.000000    99.000000  1493.000000   199.000000

       MntMeatProducts  MntFishProducts  MntSweetProducts  MntGoldProds  \
count      2205.000000      2205.000000       2205.000000   2205.000000
mean        165.312018        37.756463         27.128345     44.057143
std         217.784507        54.824635         41.130468     51.736211
min           0.000000         0.000000          0.000000      0.000000
25%          16.000000         3.000000          1.000000      9.000000
50%          68.000000        12.000000          8.000000     25.000000
75%         232.000000        50.000000         34.000000     56.000000
max        1725.000000       259.000000        262.000000    321.000000

       NumDealsPurchases  NumWebPurchases  NumCatalogPurchases  \
count        2205.000000      2205.000000          2205.000000
mean            2.318367         4.100680             2.645351
std             1.886107         2.737424             2.798647
min             0.000000         0.000000             0.000000
25%             1.000000         2.000000             0.000000
50%             2.000000         4.000000             2.000000
75%             3.000000         6.000000             4.000000
max            15.000000        27.000000            28.000000

       NumStorePurchases  NumWebVisitsMonth  AcceptedCmp3  AcceptedCmp4  \
count        2205.000000        2205.000000   2205.000000   2205.000000
mean            5.823583           5.336961      0.073923      0.074376
std             3.241796           2.413535      0.261705      0.262442
min             0.000000           0.000000      0.000000      0.000000
25%             3.000000           3.000000      0.000000      0.000000
50%             5.000000           6.000000      0.000000      0.000000
75%             8.000000           7.000000      0.000000      0.000000
max            13.000000          20.000000      1.000000      1.000000

       AcceptedCmp5  AcceptedCmp1  AcceptedCmp2     Complain  Z_CostContact  \
```

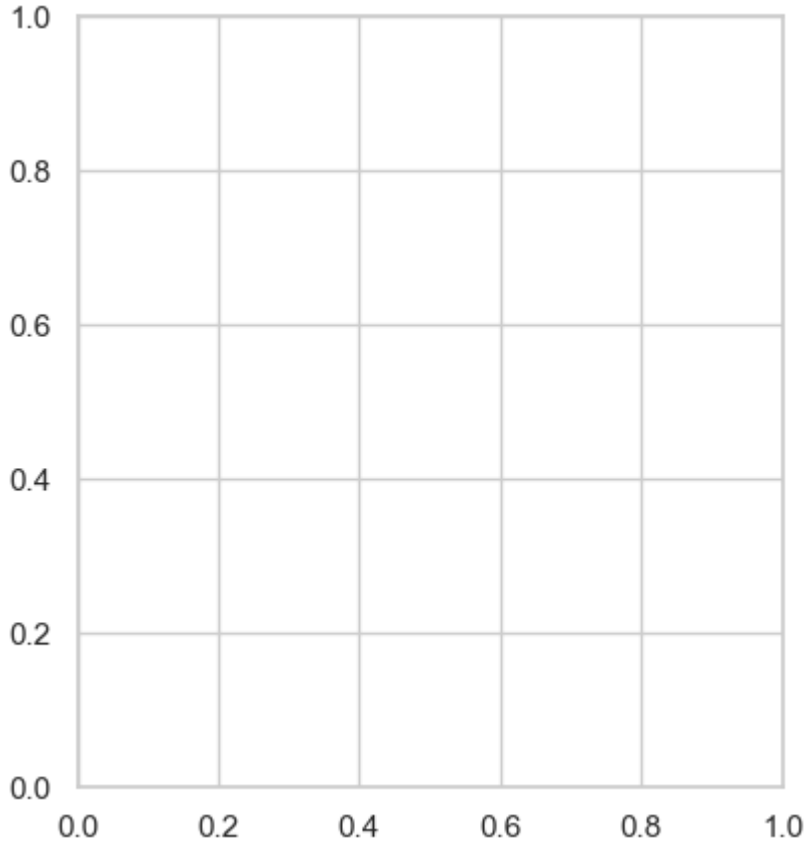
```
count   2205.000000   2205.000000   2205.000000  2205.000000         2205.0
mean       0.073016      0.064399      0.013605     0.009070            3.0
std        0.260222      0.245518      0.115872     0.094827            0.0
min        0.000000      0.000000      0.000000     0.000000            3.0
25%        0.000000      0.000000      0.000000     0.000000            3.0
50%        0.000000      0.000000      0.000000     0.000000            3.0
75%        0.000000      0.000000      0.000000     0.000000            3.0
max        1.000000      1.000000      1.000000     1.000000            3.0

       Z_Revenue    Response          Age  Customer_Days  marital_Divorced  \
count     2205.0  2205.00000  2205.000000    2205.000000       2205.000000
mean        11.0     0.15102    51.095692    2512.718367          0.104308
std          0.0     0.35815    11.705801     202.563647          0.305730
min         11.0     0.00000    24.000000    2159.000000          0.000000
25%         11.0     0.00000    43.000000    2339.000000          0.000000
50%         11.0     0.00000    50.000000    2515.000000          0.000000
75%         11.0     0.00000    61.000000    2688.000000          0.000000
max         11.0     1.00000    80.000000    2858.000000          1.000000

       marital_Married  marital_Single  marital_Together  marital_Widow  \
count      2205.000000     2205.000000       2205.000000    2205.000000
mean          0.387302        0.216327          0.257596       0.034467
std           0.487244        0.411833          0.437410       0.182467
min           0.000000        0.000000          0.000000       0.000000
25%           0.000000        0.000000          0.000000       0.000000
50%           0.000000        0.000000          0.000000       0.000000
75%           1.000000        0.000000          1.000000       0.000000
max           1.000000        1.000000          1.000000       1.000000

       education_2n Cycle  education_Basic  education_Graduation  \
count         2205.000000      2205.000000           2205.000000
mean             0.089796         0.024490              0.504762
std              0.285954         0.154599              0.500091
min              0.000000         0.000000              0.000000
25%              0.000000         0.000000              0.000000
50%              0.000000         0.000000              1.000000
75%              0.000000         0.000000              1.000000
max              1.000000         1.000000              1.000000

       education_Master  education_PhD     MntTotal  MntRegularProds  \
count       2205.000000    2205.000000  2205.000000      2205.000000
```

mean	0.165079	0.215873	562.764626	518.707483
std	0.371336	0.411520	575.936911	553.847248
min	0.000000	0.000000	4.000000	-283.000000
25%	0.000000	0.000000	56.000000	42.000000
50%	0.000000	0.000000	343.000000	288.000000
75%	0.000000	0.000000	964.000000	884.000000
max	1.000000	1.000000	2491.000000	2458.000000

	AcceptedCmpOverall
count	2205.00000
mean	0.29932
std	0.68044
min	0.00000
25%	0.00000
50%	0.00000
75%	0.00000
max	4.00000

```
---------------------------------------------------------------------------
KeyError                                  Traceback (most recent call last)
File ~\AppData\Local\Programs\Python\Python312\Lib\site-packages\pandas\core\indexes\base.py:3805, in Index.get_loc(self, key)
   3804 try:
-> 3805     return self._engine.get_loc(casted_key)
   3806 except KeyError as err:

File index.pyx:167, in pandas._libs.index.IndexEngine.get_loc()

File index.pyx:196, in pandas._libs.index.IndexEngine.get_loc()

File pandas\\_libs\\hashtable_class_helper.pxi:7081, in pandas._libs.hashtable.PyObjectHashTable.get_item()

File pandas\\_libs\\hashtable_class_helper.pxi:7089, in pandas._libs.hashtable.PyObjectHashTable.get_item()

KeyError: 'Income'

The above exception was the direct cause of the following exception:

KeyError                                  Traceback (most recent call last)
Cell In[28], line 6
      4 plt.figure(figsize=(10, 5))
      5 plt.subplot(1, 2, 1)
----> 6 sns.histplot(df_cleaned['Income'], kde=True)
      7 plt.title('Annual Income Distribution')
      9 plt.subplot(1, 2, 2)

File ~\AppData\Local\Programs\Python\Python312\Lib\site-packages\pandas\core\frame.py:4102, in DataFrame.__getitem__(self, key)
   4100 if self.columns.nlevels > 1:
   4101     return self._getitem_multilevel(key)
-> 4102 indexer = self.columns.get_loc(key)
   4103 if is_integer(indexer):
   4104     indexer = [indexer]

File ~\AppData\Local\Programs\Python\Python312\Lib\site-packages\pandas\core\indexes\base.py:3812, in Index.get_loc(self, key)
   3807     if isinstance(casted_key, slice) or (
   3808         isinstance(casted_key, abc.Iterable)
   3809         and any(isinstance(x, slice) for x in casted_key)
   3810     ):
   3811         raise InvalidIndexError(key)
```

```
-> 3812     raise KeyError(key) from err
   3813 except TypeError:
   3814     # If we have a listlike key, _check_indexing_error will raise
   3815     #  InvalidIndexError. Otherwise we fall through and re-raise
   3816     #  the TypeError.
   3817     self._check_indexing_error(key)

KeyError: 'Income'
```

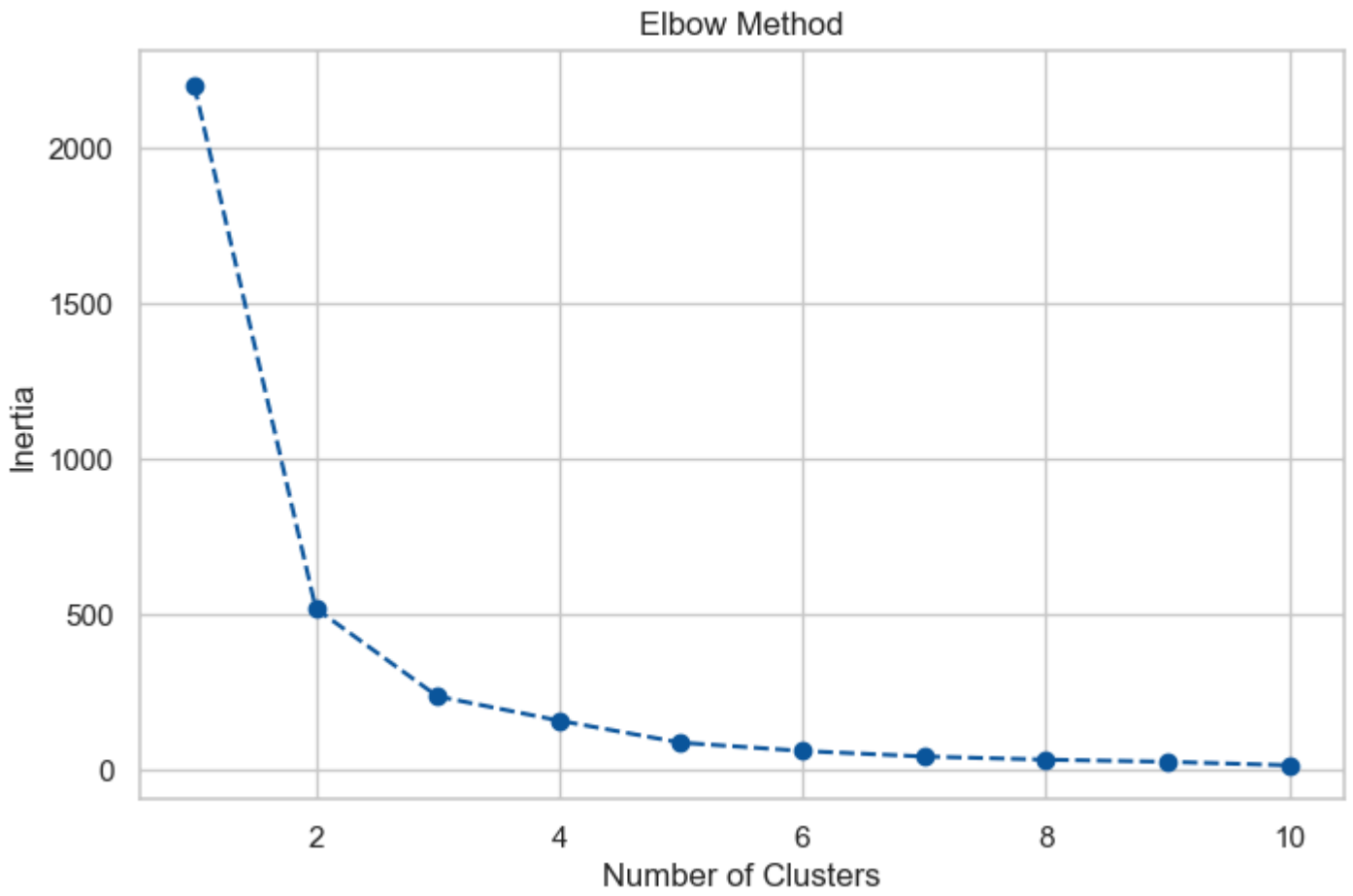


In [30]:
```
from sklearn.preprocessing import StandardScaler
```

In [32]:
```
scaler = StandardScaler()
df_scaled = scaler.fit_transform(df_cleaned[['NumStorePurchases']])
```

In [35]:
```
from sklearn.cluster import KMeans
```

In [36]:
```
inertia = []
for n in range(1, 11):
    kmeans = KMeans(n_clusters=n, init='k-means++', random_state=42)
    kmeans.fit(df_scaled)
    inertia.append(kmeans.inertia_)

plt.figure(figsize=(8, 5))
plt.plot(range(1, 11), inertia, marker='o', linestyle='--')
plt.title('Elbow Method')
plt.xlabel('Number of Clusters')
plt.ylabel('Inertia')
plt.show()

kmeans = KMeans(n_clusters=4, init='k-means++', random_state=42)
df_cleaned['Cluster'] = kmeans.fit_predict(df_scaled)
```

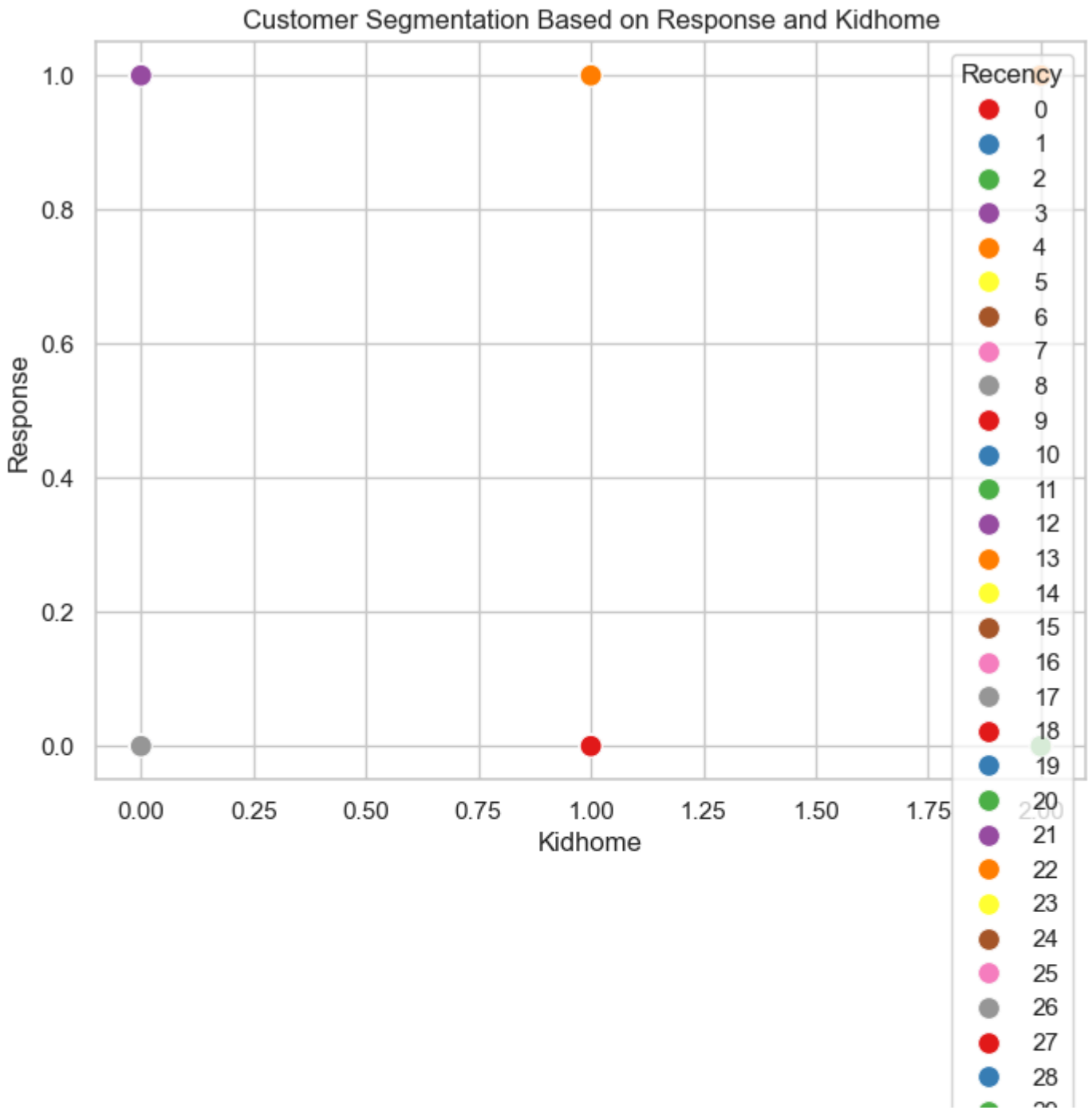


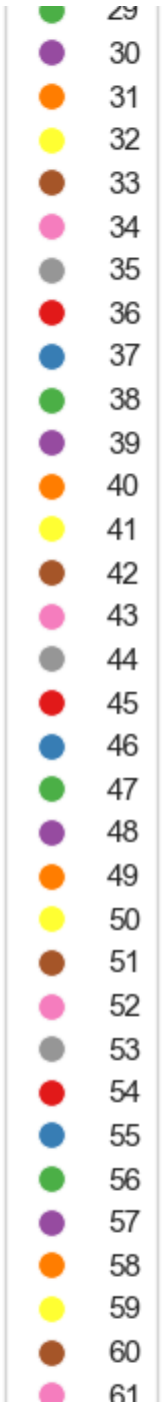
In [40]:
```
print(df_cleaned.columns)
```

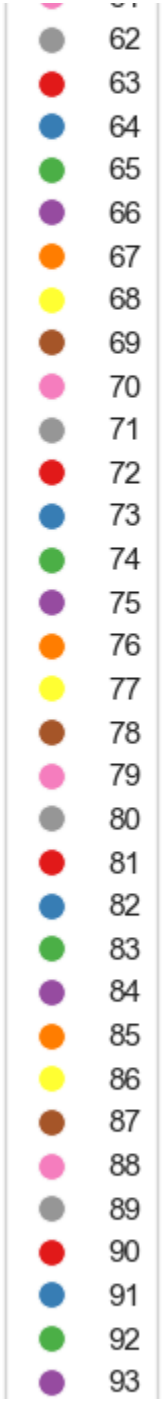
```
Index(['Kidhome', 'Teenhome', 'Recency', 'MntWines', 'MntFruits',
       'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts',
       'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',
       'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth',
       'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1',
       'AcceptedCmp2', 'Complain', 'Z_CostContact', 'Z_Revenue', 'Response',
       'Age', 'Customer_Days', 'marital_Divorced', 'marital_Married',
       'marital_Single', 'marital_Together', 'marital_Widow',
       'education_2n Cycle', 'education_Basic', 'education_Graduation',
       'education_Master', 'education_PhD', 'MntTotal', 'MntRegularProds',
       'AcceptedCmpOverall', 'Cluster'],
      dtype='object')
```

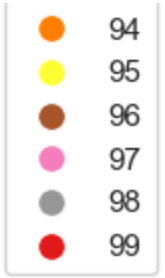
In [43]:

```
plt.figure(figsize=(8, 6))

# Use correct column names based on your dataset
sns.scatterplot(x='Kidhome', y='Response', hue='Recency', palette='Set1', data=df_cleaned, s=100)

plt.title('Customer Segmentation Based on Response and Kidhome')
plt.show()
```

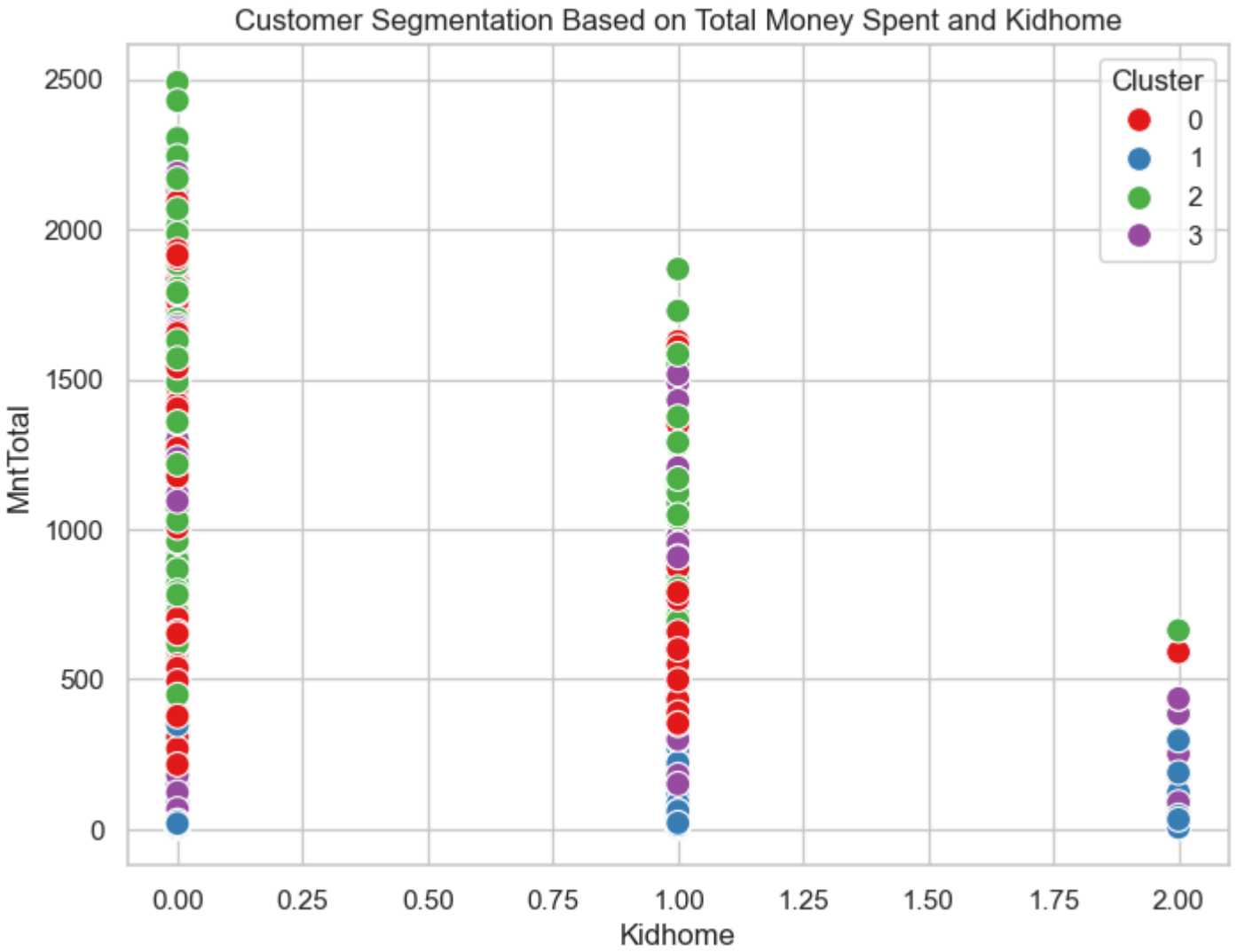

Customer Segmentation Based on Response and Kidhome
Response
Kidhome
1.0
0.8
0.6
0.4
0.2
0.0
0.00
0.25
0.50
0.75
1.00
1.25
1.50
1.75
2.00
Recency
0
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28


29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
47
48
49
50
51
52
53
54
55
56
57
58
59
60
61


62
63
64
65
66
67
68
69
70
71
72
73
74
75
76
77
78
79
80
81
82
83
84
85
86
87
88
89
90
91
92
93



In [44]:
```
plt.figure(figsize=(8, 6))

# Plot using 'Kidhome' and 'MntTotal' (total money spent)
sns.scatterplot(x='Kidhome', y='MntTotal', hue='Cluster', palette='Set1', data=df_cleaned, s=100)

plt.title('Customer Segmentation Based on Total Money Spent and Kidhome')
plt.show()
```



In [45]:
```
cluster_summary = df_cleaned.groupby('Cluster').mean()
print(cluster_summary)
```

Cluster	Kidhome	Teenhome	Recency	MntWines	MntFruits	\
0	0.191398	0.552688	46.627957	449.677419	40.236559	
1	0.829132	0.420168	48.791317	25.651261	3.596639	
2	0.094567	0.505030	50.138833	612.746479	51.358149	
3	0.466919	0.584121	50.334594	270.591682	21.580340	

Cluster	MntMeatProducts	MntFishProducts	MntSweetProducts	MntGoldProds	\
0	245.529032	54.283871	39.544086	61.690323	
1	17.338936	5.448179	4.128852	16.232493	
2	326.040241	73.175050	53.798793	69.788732	
3	143.516068	33.559546	22.200378	41.937618	

Cluster	NumDealsPurchases	NumWebPurchases	NumCatalogPurchases	\
0	2.688172	5.608602	3.651613	
1	1.802521	1.858543	0.530812	
2	2.315895	5.734406	4.859155	
3	2.691871	4.266541	2.534972	

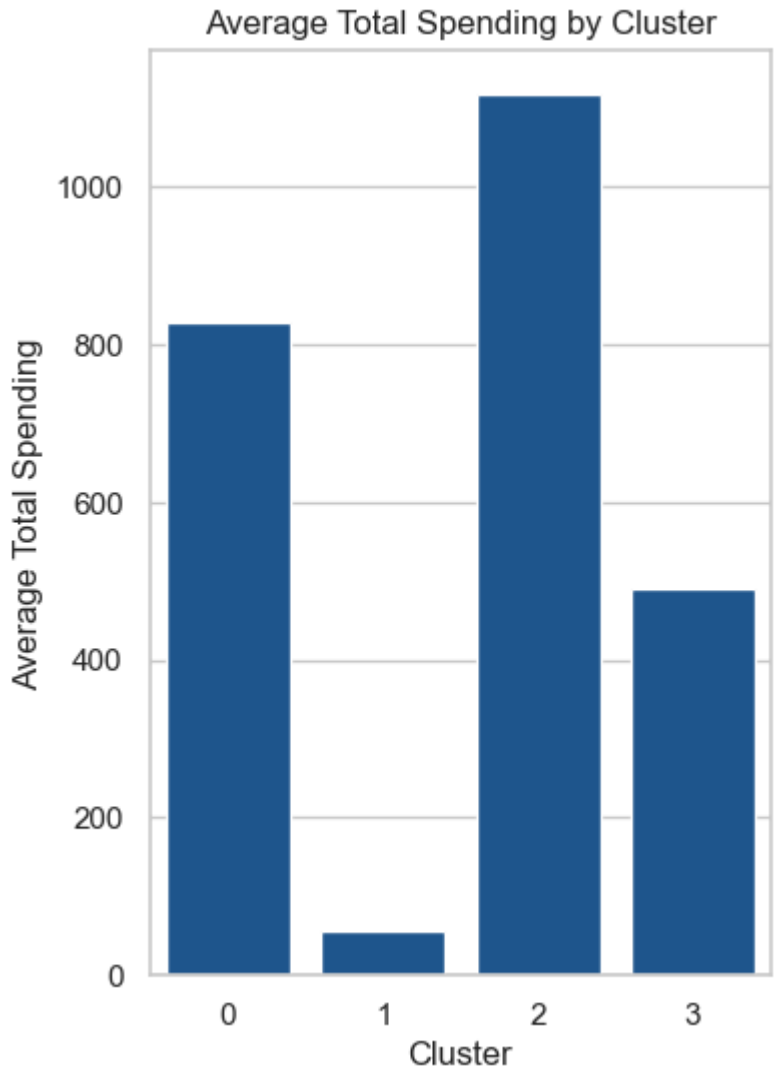
Cluster	NumStorePurchases	NumWebVisitsMonth	AcceptedCmp3	AcceptedCmp4	\
0	6.935484	4.498925	0.070968	0.135484	
1	2.644258	6.767507	0.103641	0.002801	
2	10.867203	3.943662	0.052314	0.130785	
3	4.398866	5.451796	0.056711	0.064272	

Cluster	AcceptedCmp5	AcceptedCmp1	AcceptedCmp2	Complain	Z_CostContact	\
0	0.133333	0.105376	0.019355	0.006452	3.0	
1	0.000000	0.000000	0.002801	0.014006	3.0	
2	0.140845	0.124748	0.028169	0.008048	3.0	
3	0.054820	0.058601	0.009452	0.005671	3.0	

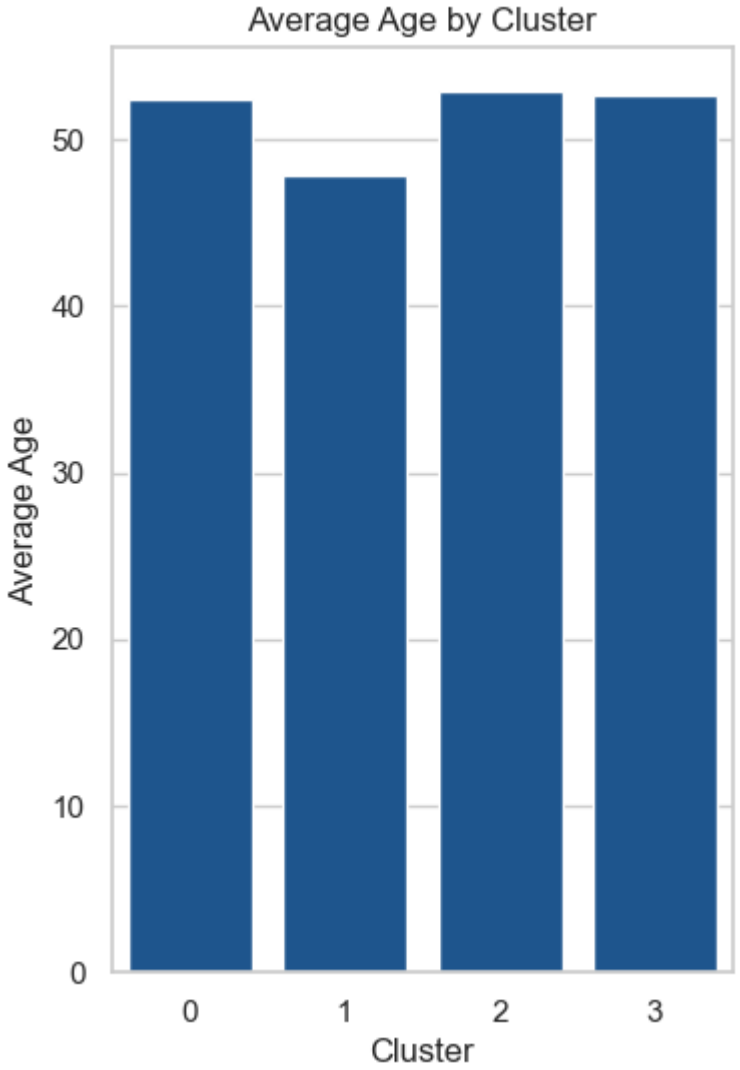
Cluster	Z_Revenue	Response	Age	Customer_Days	marital_Divorced	\
0	11.0	0.204301	52.449462	2519.612903	0.094624	
1	11.0	0.109244	47.844538	2479.571429	0.102241	
2	11.0	0.154930	52.875252	2545.945674	0.110664	
3	11.0	0.156900	52.621928	2520.179584	0.109641	

```
         marital_Married  marital_Single  marital_Together  marital_Widow  \
Cluster
0               0.384946        0.208602          0.264516       0.047312
1               0.378151        0.236695          0.256303       0.026611
2               0.396378        0.211268          0.243461       0.038229
3               0.393195        0.200378          0.266541       0.030246

         education_2n Cycle  education_Basic  education_Graduation  \
Cluster
0                  0.083871         0.002151              0.537634
1                  0.109244         0.068627              0.488796
2                  0.082495         0.000000              0.490946
3                  0.075614         0.007561              0.510397

         education_Master  education_PhD     MntTotal  MntRegularProds  \
Cluster
0                0.159140       0.217204   829.270968       767.580645
1                0.159664       0.173669    56.163866        39.931373
2                0.181087       0.245473  1117.118712      1047.329980
3                0.162571       0.243856   491.448015       449.510397

         AcceptedCmpOverall
Cluster
0                  0.464516
1                  0.109244
2                  0.476861
3                  0.243856
```

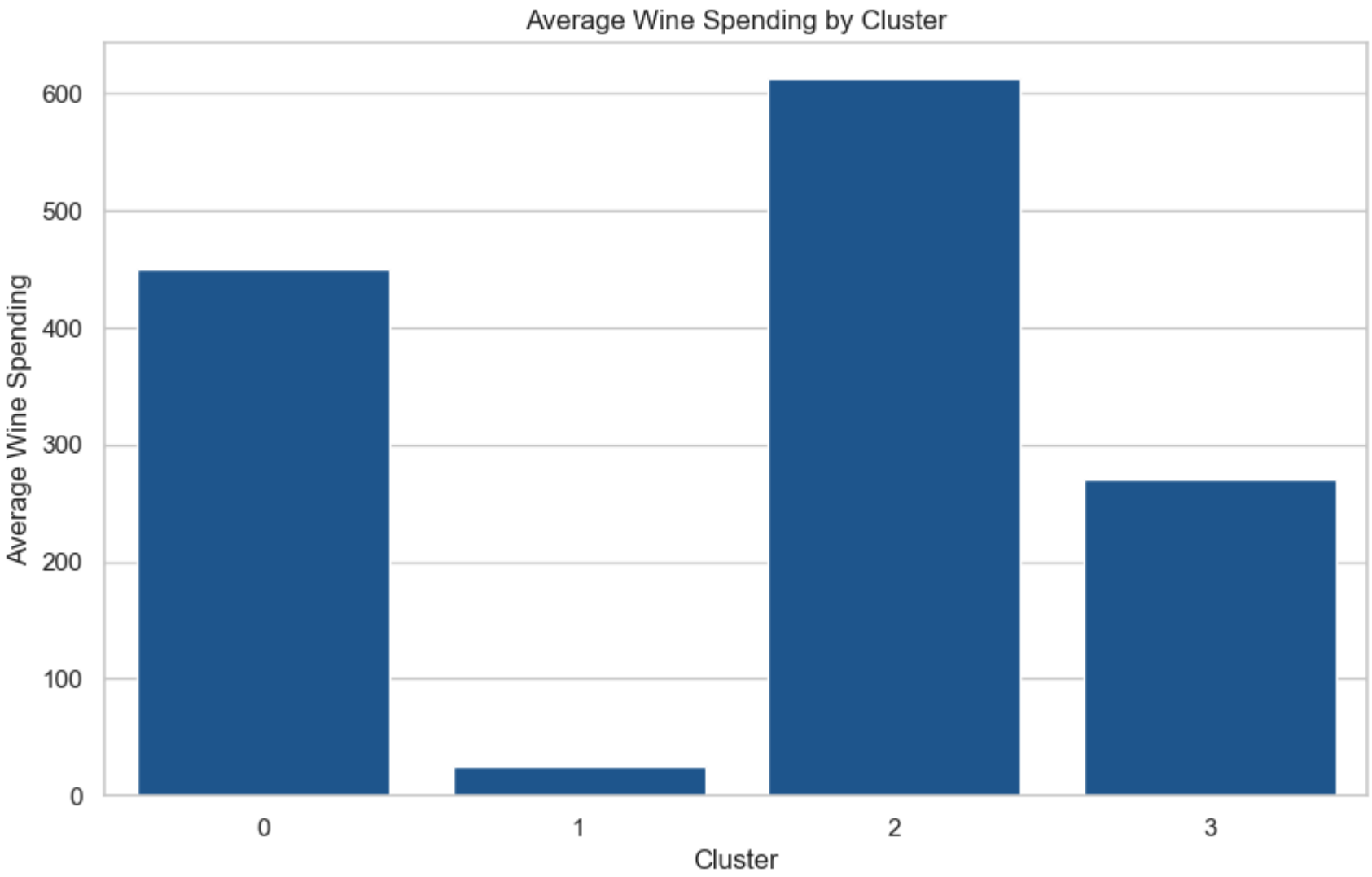
In [49]:
```
plt.figure(figsize=(4, 6))
sns.barplot(x=cluster_summary.index, y='MntTotal', data=cluster_summary)
plt.title('Average Total Spending by Cluster')
plt.ylabel('Average Total Spending')
plt.xlabel('Cluster')
plt.show()
```



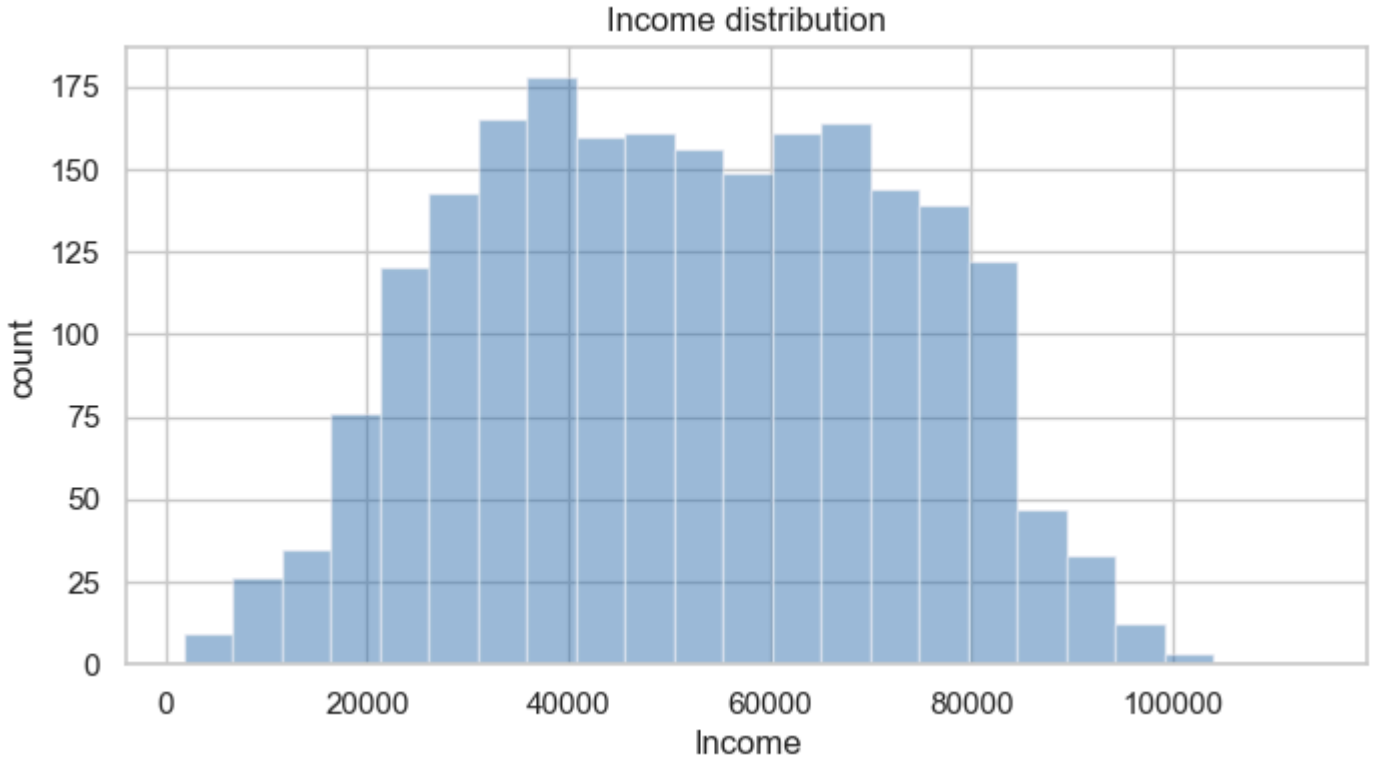
In [48]:
```
plt.figure(figsize=(4, 6))
sns.barplot(x=cluster_summary.index, y='Age', data=cluster_summary)
plt.title('Average Age by Cluster')
plt.ylabel('Average Age')
plt.xlabel('Cluster')
plt.show()
```



In [50]:
```
plt.figure(figsize=(10, 6))
sns.barplot(x=cluster_summary.index, y='MntWines', data=cluster_summary)
plt.title('Average Wine Spending by Cluster')
plt.ylabel('Average Wine Spending')
plt.xlabel('Cluster')
plt.show()
```



In [81]:
```
plt.figure(figsize=(8,4))
sns.distplot(df['Income'], kde=False, hist=True)
plt.title('Income distribution', size=12)
plt.ylabel('count');
```

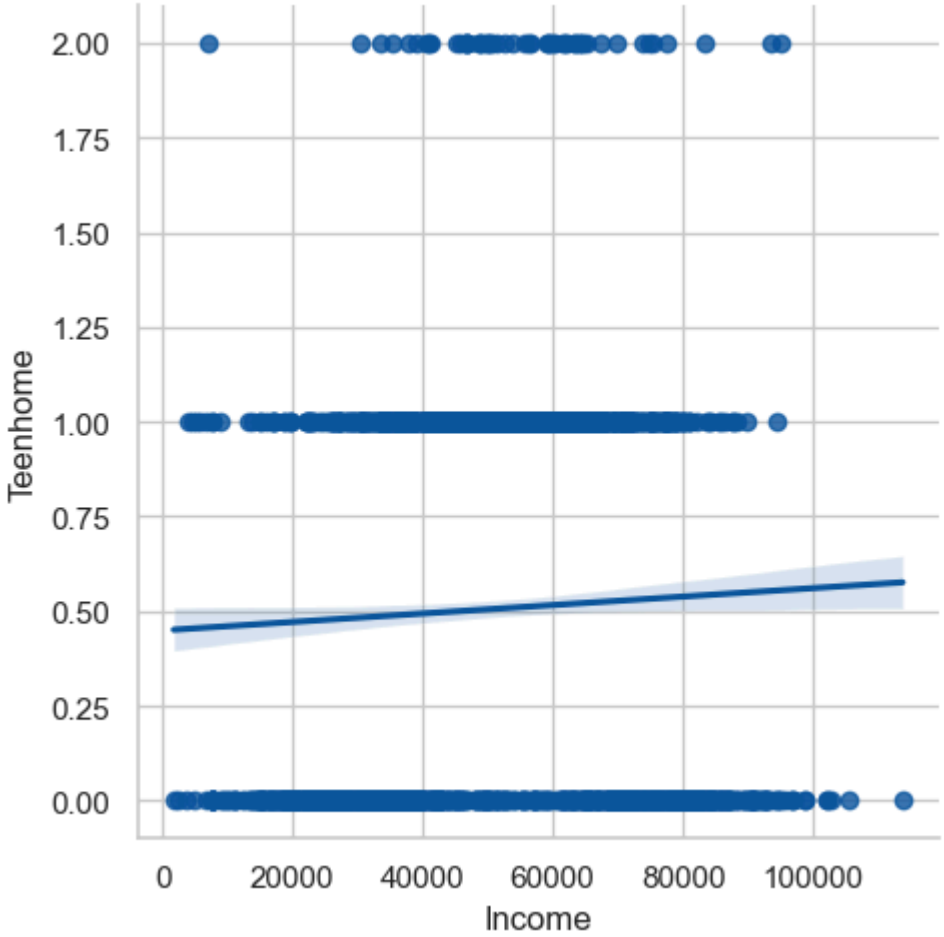


In [51]:
```
df['Income'] = df['Income'].fillna(df['Income'].median())
```

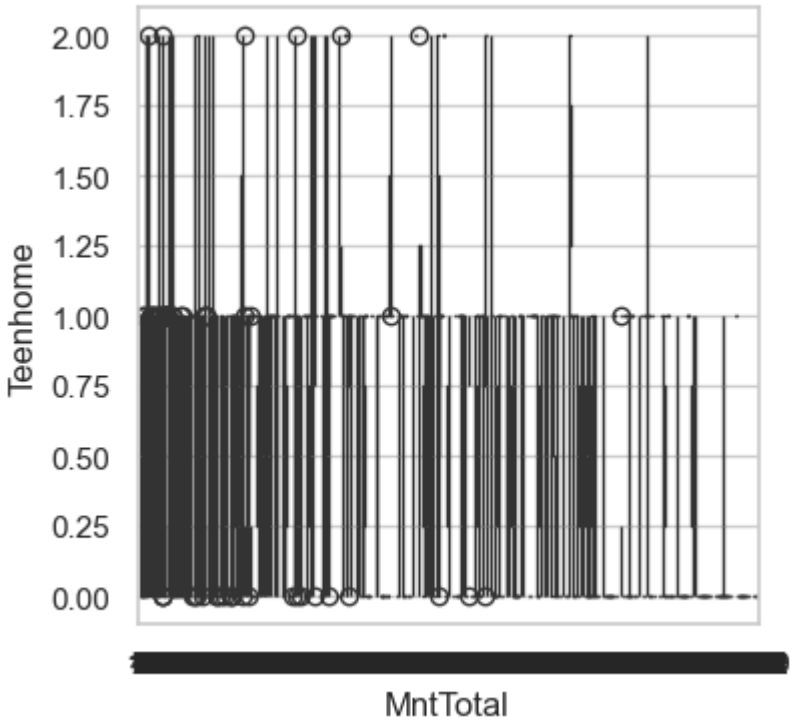
In [62]:
```
list(df.columns)
```

Out[62]:
```
['Income',
 'Kidhome',
 'Teenhome',
 'Recency',
 'MntWines',
 'MntFruits',
 'MntMeatProducts',
 'MntFishProducts',
 'MntSweetProducts',
 'MntGoldProds',
 'NumDealsPurchases',
 'NumWebPurchases',
 'NumCatalogPurchases',
 'NumStorePurchases',
 'NumWebVisitsMonth',
 'AcceptedCmp3',
 'AcceptedCmp4',
 'AcceptedCmp5',
 'AcceptedCmp1',
 'AcceptedCmp2',
 'Complain',
 'Z_CostContact',
 'Z_Revenue',
 'Response',
 'Age',
 'Customer_Days',
 'marital_Divorced',
 'marital_Married',
 'marital_Single',
 'marital_Together',
 'marital_Widow',
 'education_2n Cycle',
 'education_Basic',
 'education_Graduation',
 'education_Master',
 'education_PhD',
 'MntTotal',
 'MntRegularProds',
 'AcceptedCmpOverall',
 'Dependents']
```

In [58]:
```
sns.lmplot(x='Income', y='Teenhome', data=df[df['Income'] < 200000]);
```



In [60]:
```
plt.figure(figsize=(4,4))
sns.boxplot(x='MntTotal', y='Teenhome', data=df);
```


2.00
1.75
1.50
1.25
1.00
0.75
0.50
0.25
0.00
Teenhome
MntTotal