

```
[354.] import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
```

Data Pre-processing

Read Animal Center Intakes and Outcomes data

```
In [355.] # Read intakes csv file
intakes = pd.read_csv('processed_data/dog_intakes_unified.csv', dtype={
    'AnimalID': 'string',
    'ID': 'string',
    'Breed': 'string',
    'Color': 'string',
    'Gender': 'category',
    'Name': 'string',
    'Intake-DateTime': 'string',
    'Intake-Type': 'string',
    'Intake-Condition': 'string',
    'Intake-Age(days)': 'int',
    'parse_dates':['Intake-DateTime']}
), parse_dates=['Intake-DateTime'])
intakes.dtypes

Out[355:]
AnimalID    string[python]
ID          string[python]
Breed       string[python]
Color       string[python]
Gender      category
Name        string[python]
Intake-DateTime  datetime64[ns]
Intake-Type   string[python]
Intake-Condition string[python]
Intake-Age(days)    int32
dtype: object

In [356.] intakes

Out[356:]
AnimalID    ID          Breed      Color Gender  Name      Intake-DateTime  Intake-Type  Intake-Condition  Intake-Age(days)
0      A006100  A006100    Spinone Italiano  Yellow   Male  Scamp      2014-03-07 14:26:00  Public Assist  Normal          2190
1      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2014-12-19 10:21:00  Public Assist  Normal          2555
2      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2017-12-07 14:07:00  Stray          Normal          3650
3      A047759  A006101    Dachshund        Tricolor  Male  Oreo       2014-04-02 15:55:00  Owner Surrender  Normal          3650
4      A134067  A134067    Shetland Sheepdog  Brown    Male  Bandit     2013-11-16 09:02:00  Public Assist  Injured          12190
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
87059  A893570  A893570    Rotweiler        Black    Female  <NA>      2023-11-23 12:17:00  Stray          Normal          730
87060  A893573  A893573    Border Terrier    Brown    NaN      <NA>      2023-11-23 13:45:00  Stray          Normal          730
87061  A893578  A893578  American Staffordshire Terrier  Yellow Brindle  Male  Tiger      2023-11-23 10:19:00  Stray          Injured          240
87062  A893579  A893579  American Staffordshire Terrier  Black      Male  <NA>      2023-11-22 21:00:00  Stray          Injured          730
87063  A893585  A893585    German Shepherd  Black    Male  Unknown   2023-11-23 20:19:00  Stray          Injured          30

87064 rows x 10 columns
```

```
In [357.] # Read outcomes csv file
outcomes = pd.read_csv('processed_data/dog_outcomes_unified.csv', dtype={
    'AnimalID': 'string',
    'ID': 'string',
    'Breed': 'string',
    'Color': 'string',
    'Gender': 'category',
    'Name': 'string',
    'Date-of-Birth': 'string',
    'Outcome-DateTime': 'string',
    'Outcome-Type': 'string',
    'Outcome-SubType': 'string',
    'Outcome-Age(days)': 'int',
    'parse_dates':['Date-of-Birth', 'Outcome-DateTime']}
), parse_dates=['Date-of-Birth', 'Outcome-DateTime'])
outcomes.dtypes

Out[357:]
AnimalID    string[python]
ID          string[python]
Breed       string[python]
Color       string[python]
Gender      category
Name        string[python]
Date-of-Birth  datetime64[ns]
Outcome-DateTime  datetime64[ns]
Outcome-Type   string[python]
Outcome-SubType string[python]
Outcome-Age(days)    int32
dtype: object

In [358.] outcomes

Out[358:]
AnimalID    ID          Breed      Color Gender  Name      Date-Of-Birth  Outcome-DateTime  Outcome-Type  Outcome-SubType  Outcome-Age(days)
0      A006100  A006100    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2014-03-08 17:10:00  Return to Owner  <NA>          2435
1      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2014-12-20 16:35:00  Return to Owner  <NA>          2722
2      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2017-12-07 00:00:00  Return to Owner  <NA>          3804
3      A047759  A047759    Dachshund        Tricolor  Male  Oreo       2004-04-02  2014-04-07 15:12:00  Transfer         Partner          3658
4      A134067  A134067    Shetland Sheepdog  Brown    Male  Bandit     1997-10-16  2013-11-16 11:54:00  Return to Owner  <NA>          5875
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
86974  A893431  A893431    Chihuahua        Tricolor  Female  Chili      2015-11-21  2023-11-21 15:41:00  Return to Owner  <NA>          2923
86975  A893432  A893432    Chihuahua        Tan       Female  Coco       2015-11-21  2023-11-21 15:41:00  Return to Owner  <NA>          2923
86976  A893452  A893452    Maltese          White     Female  Sophie     2016-11-21  2023-11-22 11:26:00  Return to Owner  <NA>          2557
86977  A893529  A893529  Labrador Retriever  White     Female  <NA>       2023-09-22  2023-11-22 16:51:00  Transfer         Partner          62
86978  A893585  A893585    German Shepherd  Black    Male  Unknown   2023-09-24  2023-11-24 14:12:00  Transfer         Partner          62

86979 rows x 11 columns
```

Fill in missing values

```
In [359.] # Fill in missing value in Name and Outcome-Subtype with Unknown
intakes['Name'] = intakes['Name'].fillna('Unknown')
outcomes['Name'] = outcomes['Name'].fillna('Unknown')
intakes['Outcome-SubType'] = intakes['Outcome-SubType'].fillna('Unknown')
# Drop rows that still have missing value
intakes = intakes.dropna()
outcomes = outcomes.dropna()

87064 rows x 10 columns
```

Remove Abnormal values

```
In [360.] # Remove negative age values
intakes = intakes[intakes['Intake-Age(days)'] >= 0]
outcomes = outcomes[outcomes['Outcome-Age(days)'] >= 0]

In [361.] intakes

Out[361:]
AnimalID    ID          Breed      Color Gender  Name      Intake-DateTime  Intake-Type  Intake-Condition  Intake-Age(days)
0      A006100  A006100    Spinone Italiano  Yellow   Male  Scamp      2014-03-07 14:26:00  Public Assist  Normal          2190
1      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2014-12-19 10:21:00  Public Assist  Normal          2555
2      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2017-12-07 14:07:00  Stray          Normal          3650
3      A047759  A006101    Dachshund        Tricolor  Male  Oreo       2014-04-02 15:55:00  Owner Surrender  Normal          3650
4      A134067  A134067    Shetland Sheepdog  Brown    Male  Bandit     1997-10-16  2013-11-16 09:02:00  Public Assist  Injured          12190
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
86402  A893565  A893565    Chihuahua        Buff      Female  Unknown   2023-11-23 08:31:00  Stray          Injured          365
86403  A893570  A893570    Rotweiler        Black    Female  Unknown   2023-11-23 12:17:00  Stray          Injured          730
86404  A893578  A893578  American Staffordshire Terrier  Yellow Brindle  Male  Tiger      2023-11-23 20:19:00  Stray          Normal          240
86405  A893579  A893579  American Staffordshire Terrier  Black      Male  Unknown   2023-11-23 21:00:00  Stray          Injured          730
86406  A893585  A893585    German Shepherd  Black    Male  Unknown   2023-11-23 20:19:00  Stray          Injured          30

86396 rows x 10 columns
```

```
In [362.] outcomes

Out[362:]
AnimalID    ID          Breed      Color Gender  Name      Date-Of-Birth  Outcome-DateTime  Outcome-Type  Outcome-SubType  Outcome-Age(days)
0      A006100  A006100    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2014-03-08 17:10:00  Return to Owner  Unknown          2435
1      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2014-12-20 16:35:00  Return to Owner  Unknown          2722
2      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2017-12-07 00:00:00  Return to Owner  Unknown          3804
3      A047759  A047759    Dachshund        Tricolor  Male  Oreo       2004-04-02  2014-04-07 15:12:00  Transfer         Partner          3658
4      A134067  A134067    Shetland Sheepdog  Brown    Male  Bandit     1997-10-16  2013-11-16 11:54:00  Return to Owner  Unknown          5875
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
86301  A893431  A893431    Chihuahua        Tricolor  Female  Chili      2015-11-21  2023-11-21 15:41:00  Return to Owner  Unknown          2923
86302  A893432  A893432    Chihuahua        Tan       Female  Coco       2015-11-21  2023-11-21 15:41:00  Return to Owner  Unknown          2923
86303  A893452  A893452    Maltese          White     Female  Sophie     2016-11-21  2023-11-22 11:26:00  Return to Owner  Unknown          2557
86304  A893529  A893529  Labrador Retriever  White     Female  Unknown   2023-09-22  2023-11-22 16:51:00  Transfer         Partner          62
86305  A893585  A893585    German Shepherd  Black    Male  Unknown   2023-09-24  2023-11-24 14:12:00  Transfer         Partner          62

86285 rows x 11 columns

Add a new column of age category according to the outcomes age
```

```
In [363.] conditions = [
    (outcomes['Outcome-Age(days)'] <= 185),
    (outcomes['Outcome-Age(days)'] >= 185) & (outcomes['Outcome-Age(days)'] <= 738),
    (outcomes['Outcome-Age(days)'] > 738) & (outcomes['Outcome-Age(days)'] <= 2555),
    (outcomes['Outcome-Age(days)'] > 2555)
]
values = ['Baby', 'Young', 'Adult', 'Senior']
outcomes['Age'] = np.select(conditions, values)

Add a new boolean column 'IsAdopted' according to 'Outcome-type'
```

```
In [364.] # Define adopted as adoption outcome-type
adopted = ['Adoption']
outcomes['IsAdopted'] = np.where(outcomes['Outcome-Type'].isin(adopted), True, False)

In [365.] outcomes

Out[365:]
AnimalID    ID          Breed      Color Gender  Name      Date-Of-Birth  Outcome-DateTime  Outcome-Type  Outcome-SubType  Outcome-Age(days)  Age  IsAdopted
0      A006100  A006100    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2014-03-08 17:10:00  Return to Owner  Unknown          2435  Adult  False
1      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2014-12-20 16:35:00  Return to Owner  Unknown          2722  Senior  False
2      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2007-07-09  2017-12-07 00:00:00  Return to Owner  Unknown          3804  Senior  False
3      A047759  A047759    Dachshund        Tricolor  Male  Oreo       2004-04-02  2014-04-07 15:12:00  Transfer         Partner          3658  Senior  False
4      A134067  A134067    Shetland Sheepdog  Brown    Male  Bandit     1997-10-16  2013-11-16 11:54:00  Return to Owner  Unknown          5875  Senior  False
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
86301  A893431  A893431    Chihuahua        Tricolor  Female  Chili      2015-11-21  2023-11-21 15:41:00  Return to Owner  Unknown          2923  Senior  False
86302  A893432  A893432    Chihuahua        Tan       Female  Coco       2015-11-21  2023-11-21 15:41:00  Return to Owner  Unknown          2923  Senior  False
86303  A893452  A893452    Maltese          White     Female  Sophie     2016-11-21  2023-11-22 11:26:00  Return to Owner  Unknown          2557  Senior  False
86304  A893529  A893529  Labrador Retriever  White     Female  Unknown   2023-09-22  2023-11-22 16:51:00  Transfer         Partner          62    Baby   False
86305  A893585  A893585    German Shepherd  Black    Male  Unknown   2023-09-24  2023-11-24 14:12:00  Transfer         Partner          62    Baby   False

86285 rows x 13 columns
```

Merge intakes and outcomes dataframes by ID

```
In [366.] intakes_outcomes = pd.merge(intakes, outcomes, how='inner', on=['AnimalID', 'ID', 'Breed', 'Color', 'Gender', 'Name'], validate='1:1')
intakes_outcomes['Outcome-Age(days)'] = intakes_outcomes['Outcome-Age(days)'].astype('Int64')
intakes_outcomes

Out[366:]
AnimalID    ID          Breed      Color Gender  Name      Intake-DateTime  Intake-Type  Intake-Condition  Intake-Age(days)  Date-Of-Birth  Outcome-DateTime  Outcome-Type  Outcome-SubType  Outcome-Age(days)  Age  IsAdopted
0      A006100  A006100    Spinone Italiano  Yellow   Male  Scamp      2014-03-07 14:26:00  Public Assist  Normal          2190  2007-07-09  2014-03-08 17:10:00  Return to Owner  Unknown          2435  Adult  False
1      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2014-12-19 10:21:00  Public Assist  Normal          2555  2007-07-09  2014-12-20 16:35:00  Return to Owner  Unknown          2722  Senior  False
2      A006100  A006100+    Spinone Italiano  Yellow   Male  Scamp      2017-12-07 14:07:00  Stray          Normal          3650  2007-07-09  2017-12-07 00:00:00  Return to Owner  Unknown          3804  Senior  False
3      A134067  A134067    Shetland Sheepdog  Brown    Male  Bandit     2013-11-16 09:02:00  Public Assist  Injured          12190  1997-10-16  2013-11-16 11:54:00  Return to Owner  Unknown          5875  Senior  False
4      A141142  A141142    Labrador Retriever  Black    Female  Bettie     2013-11-16 14:46:00  Stray          Aged            11825  1998-06-01  2013-11-17 11:48:00  Return to Owner  Unknown          5648  Senior  False
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
85791  A893431  A893431    Chihuahua        Tricolor  Female  Chili      2023-11-21 12:21:00  Public Assist  Normal          2920  2015-11-21  2023-11-21 15:41:00  Return to Owner  Unknown          2923  Senior  False
85792  A893432  A893432    Chihuahua        Tan       Female  Coco       2023-11-21 12:21:00  Public Assist  Normal          2920  2015-11-21  2023-11-21 15:41:00  Return to Owner  Unknown          2923  Senior  False
85793  A893452  A893452    Maltese          White     Female  Sophie     2023-11-21 13:38:00  Public Assist  Normal          2555  2016-11-21  2023-11-22 11:26:00  Return to Owner  Unknown          2557  Senior  False
85794  A893529  A893529  Labrador Retriever  White     Female  Unknown   2023-11-22 14:26:00  Owner Surrender  Normal          30    2023-09-22  2023-11-22 16:51:00  Transfer         Partner          62    Baby   False
85795  A893585  A893585    German Shepherd  Black    Male  Unknown   2023-11-23 20:19:00  Stray          Injured          30    2023-09-24  2023-11-24 14:12:00  Transfer         Partner          62    Baby   False

85796 rows x 17 columns
```

Read dog breed data

```
In [367.] breeds = pd.read_csv('processed_data/breed_data.csv', dtype={
    'Breed': 'string',
    'Category': 'category',
    'Intelligence-Ranking': 'float64',
    'Intelligence-Category': 'category',
    'Size-Category': 'category',
    'Longevity': 'float64',
    'Total-Cost($)': 'float64',
    'Purchase-Cost($)': 'float64',
    'Food-Cost($)': 'float64'
})
breeds.dtypes

Out[367:]
Breed      string[python]
Category    category
Intelligence-Ranking  float64
Intelligence-Category category
Size-Category  category
Longevity      float64
Total-Cost($)  float64
Purchase-Cost($) float64
Food-Cost($)   float64
dtype: object

In [368.] breeds

Out[368:]
Breed      Category  Intelligence-Ranking  Intelligence-Category  Size-Category  Longevity  Total-Cost($)  Purchase-Cost($)  Food-Cost($)
0      Afterpischer  Toy          37.00      Above-Average      Small      11.42      15835.00      510.00      3180.00
1      Afghan Hound  Hound          80.00      Low          Large      11.92      20818.00      890.00      7260.00
2      Airedale Terrier  Terrier          29.00      Above-Average      Medium      11.45      16605.00      733.00      <NA>
3      Akita          Working          54.00      Average          Large      10.16      18217.00      1202.00      6188.00
4      Alaskan Malamute  Working          50.00      Average          Large      10.67      19069.00      1210.00      6499.00
...      ...      ...      ...      ...      ...      ...      ...      ...
167     Whippet       Hound          51.00      Average          Medium      12.87      18160.00      915.00      3584.00
168     Wire Fox Terrier  Terrier          51.00      Average          Small      13.17      18078.44      668.00      <NA>
169     Wirehaired Pointing Griffon  Sporting          46.00      Average          Medium      8.80      16668.00      755.00      <NA>
170     Xoloiscurdini  Non-Sporting      <NA>      <NA>      Medium      <NA>      <NA>      717.00      <NA>
171     Yorkshire Terrier  Toy          27.00      Above-Average      Small      12.60      17944.00      1057.00      3508.00

172 rows x 9 columns

Fill in missing values
```

```
In [369.] # Fill in missing values in Intelligence-Ranking and Longevity with the average value of the same category
breeds['Intelligence-Ranking'].fillna(
    breeds.groupby('Category')['Intelligence-Ranking'].transform("mean"),inplace=True
)
breeds['Intelligence-Ranking'].fillna(
    breeds['Intelligence-Ranking'].mean(),inplace=True
)

breeds['Longevity'].fillna(
    breeds.groupby('Category')['Longevity'].transform("mean"),inplace=True
)
breeds['Longevity'].fillna(
    breeds['Longevity'].mean(),inplace=True
)

In [370.] # Fill in numeric missing values in Cost with average value of the same category & size combination
breeds['Total-Cost($)'].fillna(
    breeds.groupby(['Category', 'Size-Category'])['Total-Cost($)'].transform("mean"),inplace=True
)
breeds['Total-Cost($)'].fillna(
    breeds['Total-Cost($)'].mean(),inplace=True
)

breeds['Food-Cost($)'].fillna(
    breeds.groupby(['Category', 'Size-Category'])['Food-Cost($)'].transform("mean"),inplace=True
)
breeds['Food-Cost($)'].fillna(
    breeds['Food-Cost($)'].mean(),inplace=True
)

breeds['Purchase-Cost($)'].fillna(
    breeds.groupby(['Category', 'Longevity', 'Size-Category'])['Purchase-Cost($)'].transform("mean"),inplace=True
)
breeds['Purchase-Cost($)'].fillna(
    breeds['Purchase-Cost($)'].mean(),inplace=True
)
breeds

Out[370:]
Breed      Category  Intelligence-Ranking  Intelligence-Category  Size-Category  Longevity  Total-Cost($)  Purchase-Cost($)  Food-Cost($)
0      Afterpischer  Toy          37.00      Above-Average      Small      11.42      15835.00      510.00      3180.00
1      Afghan Hound  Hound          80.00      Low          Large      11.92      20818.00      890.00      7260.00
2      Airedale Terrier  Terrier          29.00      Above-Average      Medium      11.45      16605.00      733.00      4224.00
3      Akita          Working          54.00      Average          Large      10.16      18217.00      1202.00      6188.00
4      Alaskan Malamute  Working          50.00      Average          Large      10.67      19069.00      1210.00      6499.00
...      ...      ...      ...      ...      ...      ...      ...      ...
167     Whippet       Hound          51.00      Average          Medium      12.87      18160.00      915.00      3584.00
168     Wire Fox Terrier  Terrier          51.00      Average          Small      13.17      18078.44      668.00      <NA>
169     Wirehaired Pointing Griffon  Sporting          46.00      Average          Medium      8.80      16668.00      755.00      4219.45
170     Xoloiscurdini  Non-Sporting      46.71      NaN          Medium      10.98      15294.33      717.00      3790.29
171     Yorkshire Terrier  Toy          27.00      Above-Average      Small      12.60      17944.00      1057.00      3508.00

172 rows x 9 columns
```

```
In [371.] # Cast object numbers to float
pd.options.display.float_format = '{:2f}'.format
breeds['Longevity'] = breeds['Longevity'].astype('float')
# Cast object numbers to int
breeds['Intelligence-Ranking'] = breeds['Intelligence-Ranking'].astype('int')
breeds['Total-Cost($)'] = breeds['Total-Cost($)'].astype('int')
breeds['Purchase-Cost($)'] = breeds['Purchase-Cost($)'].astype('int')
breeds['Food-Cost($)'] = breeds['Food-Cost($)'].astype('int')
breeds.dtypes

Out[371:]
Breed      string[python]
Category    category
Intelligence-Ranking  int32
Intelligence-Category category
Size-Category  category
Longevity      float64
Total-Cost($)  int32
Purchase-Cost($) int32
Food-Cost($)   int32
dtype: object

In [372.] # Fill in missing values in Intelligence-Category according to Intelligence-Ranking
intel_conditions = [
    (breeds['Intelligence-Ranking'] <= 10),
    (breeds['Intelligence-Ranking'] > 10) & (breeds['Intelligence-Ranking'] <= 26),
    (breeds['Intelligence-Ranking'] > 26) & (breeds['Intelligence-Ranking'] <= 38),
    (breeds['Intelligence-Ranking'] > 38) & (breeds['Intelligence-Ranking'] <= 54),
    (breeds['Intelligence-Ranking'] > 54) & (breeds['Intelligence-Ranking'] <= 69),
    (breeds['Intelligence-Ranking'] > 69)
]
intel_values = ['Brightest', 'Excellent', 'Above-Average', 'Average', 'Fair', 'Low']
breeds['Intelligence-Category'] = np.select(intel_conditions, intel_values)
breeds

Out[372:]
Breed      Category  Intelligence-Ranking  Intelligence-Category  Size-Category  Longevity  Total-Cost($)  Purchase-Cost($)  Food-Cost($)  Cost-Category
0      Afterpischer  Toy          37      Above-Average      Small      11.42      15835      510      3180      Saver
1      Afghan Hound  Hound          80      Low          Large      11.92      20818      890      7260      High-End
2      Airedale Terrier  Terrier          29      Above-Average      Medium      11.45      16605      733      4224      Saver
3      Akita          Working          54      Average          Large      10.16      18217      1202      6188      Budget
4      Alaskan Malamute  Working          50      Average          Large      10.67      19069      1210      6499      Budget
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
167     Whippet       Hound          51      Average          Medium      12.87      18160      915      3584      Budget
168     Wire Fox Terrier  Terrier          51      Average          Small      13.17      18078      668      3809      Budget
169     Wirehaired Pointing Griffon  Sporting          46      Average          Medium      8.80      16668      755      4219      Saver
170     Xoloiscurdini  Non-Sporting      46      Average          Medium      10.98      15294      717      3790      Saver
171     Yorkshire Terrier  Toy          27      Above-Average      Small      12.60      17844      1057      3508      Budget

172 rows x 10 columns
```

Use Agglomerative Clustering to add a new column 'Cost-Category' based on Total-Cost, Purchase-Cost, Food-Cost

```
In [373.] from sklearn.cluster import AgglomerativeClustering
import scipy.cluster.hierarchy as shc

In [374.] cost_df = breeds[['Total-Cost($)', 'Purchase-Cost($)', 'Food-Cost($)']]
cost_df

Out[374:]
Total-Cost($)  Purchase-Cost($)  Food-Cost($)
0      15835          510      3180
1      20818          890      7260
2      16605          733      4224
3      18217          1202      6188
4      19069          1210      6499
...      ...      ...      ...
167     18160          915      3584
168     18078          668      3809
169     16668          755      4219
170     15294          717      3790
171     17944          1057      3508

172 rows x 3 columns

In [375.] dend_ward = shc.dendrogram(shc.linkage(cost_df, method='ward'))

Out[375:]
clustering_ward = shc.dendrogram(shc.linkage(cost_df, method='ward'))
clustering_ward = AgglomerativeClustering(linkage=ward, n_clusters=4)
clustering_ward.fit(cost_df)
current['Cost-Category'] = clustering_ward.labels

current['labels'] = [0, 1, 2, 3]
desired_labels = ['Saver', 'High-End', 'Mid-Price', 'High-End']
map_dict = dict(zip(current.labels, desired_labels))
breeds['cost-category'] = breeds['cost-category'].map(map_dict)
breeds

Out[375:]
Breed      Category  Intelligence-Ranking  Intelligence-Category  Size-Category  Longevity  Total-Cost($)  Purchase-Cost($)  Food-Cost($)  Cost-Category
0      Afterpischer  Toy          37      Above-Average      Small      11.42      15835      510      3180      Saver
1      Afghan Hound  Hound          80      Low          Large      11.92      20818      890      7260      High-End
2      Airedale Terrier  Terrier          29      Above-Average      Medium      11.45      16605      733      4224      Saver
3      Akita          Working          54      Average          Large      10.16      18217      1202      6188      Budget
4      Alaskan Malamute  Working          50      Average          Large      10.67      19069      1210      6499      Budget
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
167     Whippet       Hound          51      Average          Medium      12.87      18160      915      3584      Budget
168     Wire Fox Terrier  Terrier          51      Average          Small      13.17      18078      668      3809      Budget
169     Wirehaired Pointing Griffon  Sporting          46      Average          Medium      8.80      16668      755      4219      Saver
170     Xoloiscurdini  Non-Sporting      46      Average          Medium      10.98      15294      717      3790      Saver
171     Yorkshire Terrier  Toy          27      Above-Average      Small      12.60      17844      1057      3508      Budget

172 rows x 10 columns
```

```
In [376.] # Clustering - ward linkage
# Choose number of clusters as 4 according to dendrogram
clustering_ward = AgglomerativeClustering(linkage=ward, n_clusters=4)
clustering_ward.fit(cost_df)
current['Cost-Category'] = clustering_ward.labels

current['labels'] = [0, 1, 2, 3]
desired_labels = ['Saver', 'High-End', 'Mid-Price', 'High-End']
map_dict = dict(zip(current.labels, desired_labels))
breeds['cost-category'] = breeds['cost-category'].map(map_dict)
breeds

Out[376:]
Breed      Category  Intelligence-Ranking  Intelligence-Category  Size-Category  Longevity  Total-Cost($)  Purchase-Cost($)  Food-Cost($)  Cost-Category
0      Afterpischer  Toy          37      Above-Average      Small      11.42      15835      510      3180      Saver
1      Afghan Hound  Hound          80      Low          Large      11.92      20818      890      7260      High-End
2      Airedale Terrier  Terrier          29      Above-Average      Medium      11.45      16605      733      4224      Saver
3      Akita          Working          54      Average          Large      10.16      18217      1202      6188      Budget
4      Alaskan Malamute  Working          50      Average          Large      10.67      19069      1210      6499      Budget
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
167     Whippet       Hound          51      Average          Medium      12.87      18160      915      3584      Budget
168     Wire Fox Terrier  Terrier          51      Average          Small      13.17      18078      668      3809      Budget
169     Wirehaired Pointing Griffon  Sporting          46      Average          Medium      8.80      16668      755      4219      Saver
170     Xoloiscurdini  Non-Sporting      46      Average          Medium      10.98      15294      717      3790      Saver
171     Yorkshire Terrier  Toy          27      Above-Average      Small      12.60      17844      1057      3508      Budget

172 rows x 10 columns
```

```
In [377.] fig = plt.figure(figsize=(30, 7))
ax = plt.axes(projection='3d')
sc = ax.scatter3d(cost_df['Purchase-Cost($)'], cost_df['Food-Cost($)'], cost_df['Total-Cost($)'], c = clustering_ward.labels_)
ax.set_xlabel('Purchase-Cost($)')
ax.set_ylabel('Food-Cost($)')
ax.set_zlabel('Total-Cost($)')
sc = ax.legend(
    scatter.legend_elements()[0], desired_labels, loc='lower right', title='Classes'
)
plt.show()

In [378.] petfinder = pd.merge(petfinder, breeds, how='left', on=['Breed'])
petfinder

Out[378:]
Petfinder_ID  Name  Size  Gender  Age  Color  Breed  Category  Intelligence-Ranking  Intelligence-Ranking  Size-Category  Longevity  Total-Cost($)  Purchase-Cost($)  Food-Cost($)  Cost-Category
0      65653819  Elsie  Large  Female  Adult  Gray  Bull Terrier  Sporting          66      Brightest      Medium      12.04      16051      815      4017      Saver
1      69226034  Penny  Medium  Female  Baby  Black  Labrador Retriever  Sporting          67      Fair          Small      16.50      22640      588      4504      High-End
2      69313950  Pirate  Extra Large  Male  Baby  Tricolor  Saint Bernard  Working          65      Fair          Large      7.78      17336      875      8124      Saver
3      69340682  Fiona  Small  Female  Senior  Apricot  Chihuahua  Herding          67      Fair          Small      16.50      22640      588      4504      High-End
4      69402276  Rory  Medium  Female  Adult  Red  Pembroke Welsh Corgi  Herding          11      Excellent      Small      12.25      19625      597      6026      Budget
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
7334  69816070  Bubba and Cadence  Small  Male  Senior  Tricolor  Border Terrier  Terrier          30      Above-Average      Small      14.00      19075      833      3898      Budget
7335  69816083  Bubba  Large  Male  Young  Black  Parson Russell Terrier  Sporting          44      Average          Small      11.48      18578      528      3809      Budget
7336 
```