

Project 3 Github Modified

April 16, 2023

0.1 Imports

```
[97]: import pandas as pd
```

0.1.1 2.1 Trees

```
[98]: Pandas dataframe called "trees" that contains the contents of the Excel file
trees = pd.read_excel("camden_trees.xlsx")
```

```
[99]: # Displaying the first few rows
trees.head()
```

```
[99]:
```

	Identifier	Number Of Trees	Site Name	Contract Area	\
0	00060053	1.0	Russell Nurseries Estate	Housing	
1	00057855	1.0	BRECKNOCK JMI (E)	Education	
2	00059953	1.0	Estate 51 Ravenshaw Street	Housing	
3	00059915	1.0	ROSARY RC JMI (E)	Education	
4	00010762	1.0	Holly Lodge Estate	Housing	

	Scientific Name	Inspection Date	Inspection Due Date	Height In Metres	\
0	Vacant Tree Pit	NaT	NaN	NaN	
1	Vacant Tree Pit	2019-07-17	2022/2023	NaN	
2	Ficus carica	NaT	NaN	5.0	
3	Betula jacquemontii	NaT	NaN	4.0	
4	Ilex x altaclarensis	2017-06-14	2020/2021	14.0	

	Spread In Metres	Diameter In Centimetres At Breast Height	Ward Code	\
0	NaN	NaN	E05000135	
1	NaN	NaN	E05000131	
2	4.0	10.0	NaN	
3	1.0	6.0	E05000135	
4	6.0	26.0	E05000137	

	Ward Name	Easting	Northing	Longitude	Latitude	\
0	Hampstead Town	527305	185240	-0.165240	51.551693	
1	Cantelowes	529923	184782	-0.127681	51.546984	
2	NaN	0	0	NaN	NaN	
3	Hampstead Town	527249	185261	-0.166051	51.551901	

```
4      Highgate    528414    186770    -0.148704    51.565198
```

```
      Location
0    (51.551693, -0.16524)
1    (51.546984, -0.127681)
2                                NaN
3    (51.551901, -0.166051)
4    (51.565198, -0.148704)
```

```
[100]: # Displaying the number of rows and columns
trees.shape
```

```
[100]: (23444, 17)
```

```
[101]: # Displaying a list of all the columns in the dataframe
trees.columns
```

```
[101]: Index(['Identifier', 'Number Of Trees', 'Site Name', 'Contract Area',
        'Scientific Name', 'Inspection Date', 'Inspection Due Date',
        'Height In Metres', 'Spread In Metres',
        'Diameter In Centimetres At Breast Height', 'Ward Code', 'Ward Name',
        'Easting', 'Northing', 'Longitude', 'Latitude', 'Location'],
        dtype='object')
```

```
[102]: # Displaying a List the data types of each column
trees.dtypes
```

```
[102]: Identifier                object
Number Of Trees                float64
Site Name                     object
Contract Area                 object
Scientific Name               object
Inspection Date               datetime64[ns]
Inspection Due Date           object
Height In Metres              float64
Spread In Metres              float64
Diameter In Centimetres At Breast Height float64
Ward Code                    object
Ward Name                    object
Easting                      int64
Northing                     int64
Longitude                    float64
Latitude                     float64
Location                     object
dtype: object
```

0.1.2 2.2 Environmental

```
[103]: # Created a Pandas dataframe called "environmental" that contains the contents
↳ of the csv file
```

```
environmental = pd.read_csv("camden_trees_environmental.csv")
```

```
[104]: # Displayed the first few rows
```

```
environmental.head()
```

```
[104]:
```

	Identifier	Maturity	Physiological Condition	Tree Set To Be Removed	\
0	00055125	Juvenile		Good	No
1	00059429	Middle aged		Fair	No
2	00018254	Mature		Fair	No
3	00027155	Mature		Fair	No
4	00041326	Juvenile		Good	No

	Removal Reason	Capital Asset Value For Amenity Trees	\
0	NaN	115.07	
1	NaN	7518.08	
2	NaN	20419.63	
3	NaN	21447.74	
4	NaN	524.30	

	Carbon Storage In Kilograms	\
0	1.6	
1	NaN	
2	426.4	
3	448.3	
4	9.9	

	Gross Carbon Sequestration Per Year In Kilograms	\
0	0.5	
1	NaN	
2	8.8	
3	9.6	
4	1.4	

	Pollution Removal Per Year In Grams
0	5.7
1	NaN
2	215.2
3	379.1
4	12.8

```
[105]: # Displayed the number of rows and columns
```

```
environmental.shape
```

```
[105]: (23415, 9)
```

```
[106]: #Displayed a list of all the columns in the dataframe
environmental.columns
```

```
[106]: Index(['Identifier', 'Maturity', 'Physiological Condition',
        'Tree Set To Be Removed', 'Removal Reason',
        'Capital Asset Value For Amenity Trees', 'Carbon Storage In Kilograms',
        'Gross Carbon Sequestration Per Year In Kilograms',
        'Pollution Removal Per Year In Grams'],
        dtype='object')
```

```
[107]: # Listed the data types of each column
environmental.dtypes
```

```
[107]: Identifier                                object
Maturity                                       object
Physiological Condition                       object
Tree Set To Be Removed                       object
Removal Reason                               object
Capital Asset Value For Amenity Trees         float64
Carbon Storage In Kilograms                   float64
Gross Carbon Sequestration Per Year In Kilograms float64
Pollution Removal Per Year In Grams          float64
dtype: object
```

0.1.3 2.3 Common and Scientific Names

```
[108]: # Created a Pandas dataframe called "names" that contains the contents of the
        ↪ json file
names = pd.read_json("https://s3.eu-west-1.amazonaws.com/course.oc-static.com/
        ↪ projects/DAN_UK_App_P3/tree_common_names.json")
```

```
[109]: # Displayed the first few rows
names.head()
```

```
[109]:
```

	Scientific Name	Common Name
0	Carpinus betulus Lucas	Hornbeam - European
1	Prunus 'Pandora'	Cherry - Ornamental
2	Tilia unidentified species	Lime
3	Rosa unidentified species	None
4	Cedrus libani	Cedar of Lebanon

```
[110]: # Displayed the number of rows and columns
names.shape
```

```
[110]: (589, 2)
```

```
[111]: # Displayed a list of all the columns in the dataframe
names.columns
```

```
[111]: Index(['Scientific Name', 'Common Name'], dtype='object')
```

```
[112]: # Listed the data types of each column
names.dtypes
```

```
[112]: Scientific Name    object
Common Name          object
dtype: object
```

0.2 Task 3: Further Inspect the Datasets

0.2.1 3.1 Further Inspect the Trees Dataset

Site Name

```
[113]: # List of values in Site Name column and their counts
trees["Site Name"].value_counts()
```

```
[113]: WATERLOW PARK (LS)          920
Alexandra & Ainsworth Estate    289
Belsize nature reserve, Russell Nursery 278
Holly Lodge Estate             272
LINCOLN'S INN FIELDS, GARDENS (LS) 193
...
GOLDINGTON CRESCENT            1
ALLCROFT ROAD                  1
WOBBURN WALK, LAND BEHIND 4-18  1
KILBURN PRIORY                 1
GOODGE PLACE                   1
Name: Site Name, Length: 1135, dtype: int64
```

Site Name is **qualitative nominal**.

```
[114]: #List of values for Contract Area column and its counts
trees["Contract Area"].value_counts()
```

```
[114]: Highways          10062
Housing            7500
Parks              4330
Education          1288
Corporate Landlord  264
Name: Contract Area, dtype: int64
```

Contract Area is **Qualitative nominal**

```
[115]: #List of values for Scientific name and its counts
trees["Scientific Name"].value_counts()
```

```
[115]: Platanus x hispanica          3340
       Tilia europaea              1468
       Acer pseudoplatanus         941
       Betula pendula              765
       Fraxinus excelsior          754
       ...
       Vacant Tree Pit (planned: Populus tremula) 1
       Liriodendron fastigiata      1
       Sequoia sempervirens         1
       Sorbus x hybrida             1
       Vacant Tree Pit (planned: Acer rubrum 'Amstrong') 1
       Name: Scientific Name, Length: 543, dtype: int64
```

Scientific name is **Qualitative nominal**

```
[116]: #List of values for Ward name and its counts
       trees["Ward Name"].value_counts()
```

```
[116]: Highgate                    2799
       St Pancras and Somers Town  1832
       Gospel Oak                  1541
       Kilburn                     1540
       Kentish Town                1463
       Haverstock                  1424
       Hampstead Town              1340
       Holborn and Covent Garden   1293
       Fortune Green               1284
       Cantelowes                  1231
       Frognal and Fitzjohns       1229
       Bloomsbury                  1008
       Regent's Park                989
       Swiss Cottage               978
       Camden Town with Primrose Hill 899
       West Hampstead              853
       King's Cross                824
       Belsize                     691
       Name: Ward Name, dtype: int64
```

Ward Name is **Qualitative nominal**

```
[117]: # Descriptive stats for the numeric columns
       trees.describe()
```

```
[117]:      Number Of Trees  Height In Metres  Spread In Metres  \
count      23422.000000      23006.000000      23006.000000
mean         1.100034         10.307029         5.997612
std          1.288041          6.325293          4.132379
```

min	0.000000	0.000000	0.000000
25%	1.000000	5.000000	3.000000
50%	1.000000	9.000000	5.000000
75%	1.000000	15.000000	8.000000
max	67.000000	127.000000	88.000000

	Diameter In Centimetres At Breast Height	Easting	Northing \
count	23005.000000	23444.000000	23444.000000
mean	32.595262	526762.518171	184085.188662
std	26.149994	25835.857212	9121.059755
min	0.000000	0.000000	0.000000
25%	12.000000	526583.750000	183665.000000
50%	27.000000	528456.500000	184690.000000
75%	46.000000	529369.000000	185481.000000
max	228.000000	531514.000000	196188.000000

	Longitude	Latitude
count	23388.000000	23388.000000
mean	-0.155145	51.545115
std	0.025058	0.013067
min	-0.261719	51.512858
25%	-0.175734	51.537472
50%	-0.148478	51.546669
75%	-0.135978	51.554032
max	-0.105858	51.650843

```
[118]: # the data types
trees.dtypes
```

```
[118]: Identifier                object
Number Of Trees                float64
Site Name                      object
Contract Area                  object
Scientific Name                object
Inspection Date                datetime64[ns]
Inspection Due Date            object
Height In Metres               float64
Spread In Metres               float64
Diameter In Centimetres At Breast Height float64
Ward Code                      object
Ward Name                      object
Easting                        int64
Northing                       int64
Longitude                      float64
Latitude                       float64
Location                       object
dtype: object
```

```
[119]: trees["Number Of Trees"].unique()
```

```
[119]: array([ 1.,  2.,  3.,  0., nan,  5.,  6.,  7., 18.,  8., 65.,  4., 10.,  
          9., 11., 50., 12., 15., 52., 40., 33., 13., 20., 67., 21., 32.,  
          24., 26., 16., 25., 51.] )
```

```
[120]: trees["Height In Metres"].unique()
```

```
[120]: array([ nan,   5. ,   4. ,  14. ,   9. ,   0. ,   2. ,   2.5,   8. ,  
          13. ,  17. ,  10. ,   3. ,  19. ,   7. ,   6. ,   1.8,  15. ,  
          12. ,   1.5,  16. ,  20. ,  21. ,  24. ,  25. ,   2.7,  18. ,  
          11. ,  26. ,  22. ,   0.5,  16.1,  27. ,  28. ,   2.3,  22.3,  
          38. ,  15.5,  29. ,  23. ,  34. ,   3.5,   2.6,   1. ,   5.5,  
          22.5,   2.2,  36. ,  31. ,  96. ,  13.6, 127. ,  14.7,   4.5,  
          30. ,  35. ,  32. ,   9.4,  11.8,  33. ,  40. ,   2.1,   6.5,  
          41. ,  23.1,  37. ,  39. ,  12.5,   7.5,  13.7,  31.6,  23.8,  
           0.2,  12.3,  15.7,   6.8,   9.3,   3.8,  24.9,  17.4,   2.8,  
           3.2,  13.9,  12.4,  10.2,  10.8,  24.5,  30.4,   9.8,  11.4,  
          23.4,   3.7,  11.6,   7.7,   8.3,   3.6,  17.5,  19.5] )
```

```
[121]: trees["Spread In Metres"].unique()
```

```
[121]: array([ nan,   4. ,   1. ,   6. ,   7. ,   0. ,   1.5 ,   5. ,   9. ,  
          10. ,   3. ,   2. ,  13. ,   2.5 ,  12. ,  15. ,  14. ,   8. ,  
          11. ,   0.6 ,   1.8 ,  17. ,  20. ,  18. ,  16. ,  22. ,  19. ,  
           5.5 ,  26. ,   1.4 ,  23. ,  21. ,  28. ,   1.2 ,   0.5 ,   3.5 ,  
          24. ,   1.3 ,   1.6 ,  30. ,   1.7 ,  27. ,   4.2 , 11.02,   0.7 ,  
           6.5 ,   4.5 ,  31. ,  25. ,   0.8 ,   0.3 ,   2.2 ,   2.8 ,   1.65,  
          29. ,  88.  ] )
```

```
[122]: trees["Diameter In Centimetres At Breast Height"].unique()
```

```
[122]: array([ nan,  10. ,   6. ,  26. ,  29. ,   5. ,   0. ,   4. ,  12. ,  
          59. ,  52. ,  23. ,  50. ,  63. ,  15. ,   3. ,  49. ,  42. ,  
          14. ,  19. ,   9. ,  70. ,  32. ,  28. ,  34. ,  20. ,  17. ,  
          27. ,  37. ,   8. ,  45. ,  18. , 119. ,   7. ,  38. ,  55. ,  
          41. ,  75. ,  31. ,  25. ,  11. ,  30. ,  43. ,  68. ,  92. ,  
          16. ,  35. ,  58. ,  72. ,  64. ,  13. ,  61. ,  69. ,  33. ,  
          47. ,  67. , 109. , 106. ,  24. ,  51. ,  40. ,  22. ,  60. ,  
          39. ,  46. ,  57. ,  21. ,  54. , 117. ,  44. ,  82. , 114. ,  
          65. ,  84. , 118. ,   1. ,  89. ,  53. ,  36. ,  93. ,  88. ,  
          86. , 163. ,  66. ,  81. ,  74. , 100. ,  48. ,  73. ,  95. ,  
          129. ,  62. ,  85. ,  56. ,   3.5, 161. ,  71. , 105. , 102. ,  
          80. ,  87. , 101. ,  76. , 113. , 108. , 160. , 132. ,   90. ,  
          145. , 130. ,   79. , 228. , 110. ,   83. ,   78. , 122. , 170. ,  
          115. ,   2. ,   77. , 107. ,   96. , 126. ,   91. , 104. , 158. ,  
          99. ,  94. ,  16.5, 127. , 151. , 103. , 112. ,   98. ,   97. ,  
          136. , 125. , 111. , 124. , 139. , 156. , 120. , 148. , 144. ,
```



```

140. , 121. , 143. , 17.5, 154. , 159. , 142. , 197. , 123. ,
149. , 155. , 191. , 131. , 147. , 162. , 116. , 152. , 153. ,
165. , 137. , 200. , 177. , 133. , 128. , 134. , 11.5, 135. ,
150. , 187. , 210. , 166. , 138. , 10.5, 206. , 141. , 209. ,
184. , 173. , 192. , 7.5, 194. , 157. , 146. , 185. ])

```

```
[123]: trees["Easting"].unique()
```

```
[123]: array([527305, 529923, 0, ..., 527733, 524398, 525944], dtype=int64)
```

```
[124]: trees["Northing"].unique()
```

```
[124]: array([185240, 184782, 0, ..., 185755, 187062, 187313], dtype=int64)
```

```
[125]: trees["Longitude"].unique()
```

```
[125]: array([-0.16524 , -0.127681, nan, ..., -0.196884, -0.204206,
-0.173397])
```

```
[126]: trees["Latitude"].unique()
```

```
[126]: array([51.551693, 51.546984, nan, ..., 51.54329 , 51.545726,
51.531863])
```

Columns:

Number of Trees is **discrete** - Values are Int with Null's - so **NOT** a float

Height in Metres is **continuous** - Values have both int and floats

Spread in Metres is **continuous** - Values have both int and floats

Diameter In Centimetres At Breast Height is **continuous** - Values have more int but a few floats

Easting is **discrete** - Values are all int

Northing is **discrete** - Values are all int

Longitude is **continuous** - values are float

Latitude is **continuous** - values are float

0.2.2 3.2 Further Inspect the Environmental Dataset

```
[127]: environmental["Maturity"].value_counts()
```

```
[127]: Mature          10225
Middle aged         7779
Juvenile            4393
Not Applicable       377
Over Mature          191
Veteran              41
```

Name: Maturity, dtype: int64

Maturity is qualitative **nominal**

```
[128]: environmental["Physiological Condition"].value_counts()
```

```
[128]: Good          12910
Fair            9183
Poor            357
Not applicable  249
Dead            236
Excellent        8
Name: Physiological Condition, dtype: int64
```

Physiological Condition is qualitative **ordinal**

```
[129]: environmental["Tree Set To Be Removed"].value_counts()
```

```
[129]: No      23331
Yes       84
Name: Tree Set To Be Removed, dtype: int64
```

Trees set to be removed is qualitative **Binary**

```
[130]: environmental["Removal Reason"].value_counts()
```

```
[130]: Dead, dying          30
Basal decay              17
Trunk decay              10
Tree defect              5
Crown die-back          5
Dog damage               3
Unsuitable location      3
Newly planted tree failure 2
Coppiced stump           1
Touching building/structure 1
Crown decay              1
Split trunk              1
Broken/split branch      1
Climber                  1
No defects - work required 1
Suppressed               1
ATRD                     1
Name: Removal Reason, dtype: int64
```

Removal reason is qualitative **nominal**

3.2.b Descriptive Stats for Numeric Type Columns

```
[131]: # Descriptive stats for all numeric columns
environmental.describe()
```

```
[131]:      Capital Asset Value For Amenity Trees  Carbon Storage In Kilograms  \
count      22982.000000      20555.000000
mean      14056.393047      467.465454
std       24803.806595      844.926555
min         0.000000         0.500000
25%       1035.650000         24.800000
50%       5443.660000        163.900000
75%      16781.420000        497.300000
max      504725.720000        6000.000000

      Gross Carbon Sequestration Per Year In Kilograms  \
count      20555.000000
mean         8.675057
std         8.681446
min         0.000000
25%         2.200000
50%         6.100000
75%        11.700000
max        53.800000

      Pollution Removal Per Year In Grams
count      20555.000000
mean       217.740316
std       306.751920
min         0.300000
25%        29.300000
50%       108.100000
75%       297.600000
max      8223.700000
```

```
[132]: # The Data Types
environmental.dtypes
```

```
[132]: Identifier      object
Maturity             object
Physiological Condition  object
Tree Set To Be Removed  object
Removal Reason        object
Capital Asset Value For Amenity Trees  float64
Carbon Storage In Kilograms  float64
Gross Carbon Sequestration Per Year In Kilograms  float64
Pollution Removal Per Year In Grams  float64
dtype: object
```

Find out if each numerical column values are really ints or floats

```
[133]: environmental["Capital Asset Value For Amenity Trees"].unique()
```

```
[133]: array([1.1507000e+02, 7.5180800e+03, 2.0419630e+04, ..., 3.3664130e+04,  
        3.6269450e+04, 1.4801215e+05])
```

```
[134]: environmental["Carbon Storage In Kilograms"].unique()
```

```
[134]: array([1.6000e+00,          nan, 4.2640e+02, ..., 4.7233e+03, 3.7305e+03,  
        4.8100e+02])
```

```
[135]: environmental["Gross Carbon Sequestration Per Year In Kilograms"].unique()
```

```
[135]: array([ 0.5,  nan,  8.8,  9.6,  1.4, 10.1,  0.8,  7.9, 24.2,  2.4,  0.9,  
        3. , 14.6,  8.1,  4.1,  1.8, 18.8, 24.4, 13.4,  4.3,  6.5,  1.3,  
        0.6,  2.5, 13.1,  2.1,  3.9, 15.9,  1. ,  6.6,  0.4,  3.4,  2.9,  
        28.7,  0.7,  1.5, 12.5, 25.4, 11.2,  9. , 23.9,  4.4, 11.5, 11. ,  
        8.4, 10.9, 10.7,  6.9,  7. ,  3.8,  3.2,  6.1,  8.6, 30.2,  3.7,  
        15. , 30.4,  7.6, 20.1, 10.2,  8.3, 39.9,  0.3,  4.9, 14.3, 13.5,  
        16.7,  8.9,  1.6,  4.2,  3.6,  4. ,  6.7,  0.1, 19.7, 24.6,  6.4,  
        5.4,  5.9, 12.2,  7.3, 13. ,  7.1, 36.9,  9.3, 18.2, 10. ,  4.7,  
        5. , 29.8, 17.8, 18.5, 17.6,  7.2,  4.8,  5.6,  5.3, 10.5, 12.9,  
        9.4,  7.8,  1.1, 19.2, 37.3,  2.7, 12.8, 17. , 15.5, 27.6, 34.2,  
        5.7,  5.8, 17.3, 20.4,  9.9, 15.6,  7.4, 11.8,  9.2,  2.6, 21.7,  
        11.3, 10.8, 29.4, 23.1, 26.5,  1.2, 10.6, 33.6, 23.6, 11.7, 17.4,  
        15.1, 16.6,  2.3, 13.3, 16.3, 18.3, 10.4, 19. ,  1.7,  3.3, 14.1,  
        5.5, 12.6, 14.5,  3.5,  8.7, 36.6, 17.7,  5.1, 21.1, 20.2,  7.7,  
        15.2,  7.5, 30.5, 13.8,  4.6, 39.2, 37.8, 13.2, 51. , 11.6,  6.8,  
        21. ,  5.2,  8. ,  4.5, 15.3, 13.7, 18. ,  6.3,  2. ,  2.8,  9.1,  
        6. , 12.4, 11.4, 18.6,  1.9, 23.5, 18.9, 26.2, 12.1, 41.7,  2.2,  
        14.7, 14.4,  9.7, 23. , 22.1, 15.7, 12. , 14.8, 19.9, 29.7, 11.1,  
        19.6, 12.7, 31.9, 33.2, 19.1,  6.2, 28.4, 27. , 17.5, 16.4, 27.7,  
        38.4, 43.9, 35.4,  8.2, 37.1, 29.5, 22.2, 17.1, 26.3, 21.2, 20.5,  
        11.9, 31.3, 14.2,  8.5, 26.8,  0.2, 16.8, 35.6, 20.8, 22.3, 19.4,  
        20. , 14. , 36.1, 20.9, 15.8, 14.9,  9.5, 28.3, 16.9,  3.1, 37.6,  
        34.7, 17.9, 35.3,  9.8, 16.5, 21.4, 19.3, 28.5, 29. , 33.5, 40.7,  
        28.6, 13.6, 27.1, 26.7, 13.9, 21.3, 23.8, 35.1, 40.5, 32.9, 18.1,  
        37. , 20.6, 35.2, 12.3, 41.8, 22.7, 21.5, 35.7, 32.4, 16.1, 31.1,  
        25.3, 26. , 23.4, 30.1, 42.1, 36.5, 10.3, 44.3, 16. , 23.2, 28.8,  
        21.9, 32. , 30.3, 15.4, 24.3, 26.9, 25.8, 28.2, 22.8, 42.3, 33.9,  
        25.6, 38.6, 18.4, 31.2, 32.7, 28. , 30.6, 42. , 27.3, 33.8, 21.6,  
        33.3, 35.9, 24. , 40.6, 42.7, 31.5, 22.5, 20.7, 33.1, 35.5, 27.5,  
        24.8, 26.4, 22.6, 18.7, 32.6, 36.4, 16.2, 38.5, 22.4, 22.9, 29.2,  
        25.9, 19.5, 24.5, 20.3, 27.9, 29.6, 25. , 27.4, 30.7, 25.1, 34.6,  
        38. , 34.4, 36.7, 34.1, 32.5, 27.8, 32.8, 43.1, 17.2, 37.5, 39.3,  
        29.1, 30.8, 34.8, 29.9, 38.3, 23.3, 21.8, 41.3, 38.7, 36.3, 30. ,  
        41.5, 37.2, 24.9, 19.8, 28.9, 42.2, 39.5, 31.8, 37.4, 24.1,  0. ,  
        37.7, 32.3, 36.8, 27.2, 24.7, 36.2, 39.7, 42.9, 49.5, 31.6, 38.9,
```

```
40.9, 44.2, 32.1, 42.5, 25.7, 35.8, 38.2, 40.8, 35. , 34. , 31.7,
40.1, 26.1, 25.2, 39. , 41.1, 33.4, 32.2, 43.3, 33.7, 26.6, 39.1,
41.4, 43.2, 31. , 41.9, 41. , 42.4, 43. , 40.3, 40.2, 29.3, 31.4,
22. , 42.8, 43.7, 23.7, 34.5, 41.6, 39.6, 39.8, 37.9, 43.8, 42.6,
41.2, 43.6, 36. , 39.4, 43.5, 25.5, 38.1, 34.9, 28.1, 43.4, 53.8,
30.9, 40.4, 34.3, 45.5, 52.7, 40. , 38.8, 44.9, 33. ])
```

```
[136]: environmental["Pollution Removal Per Year In Grams"].unique()
```

```
[136]: array([ 5.7,   nan, 215.2, ...,   8. , 399.9,  60.1])
```

Columns

Capital Asset Value For Amenity Trees is **continuous** - values are float

Carbon Storage In Kilograms is **continuous** - values are float

Gross Carbon Sequestration Per Year In Kilograms is **continuous** - values are float

Pollution Removal Per Year In Grams is **continuous** - values are float

0.2.3 3.3 Further Inspect the Common Names Dataset

3.3.a (Names) Counts of Values for String Type Columns

```
[137]: names["Scientific Name"].value_counts()
```

```
[137]: Cupressocyparis leylandii          2
      Larix decidua                      2
      Salix fragilis                     2
      Alnus cordata                     2
      Populus nigra                      2
      ..
      Pyrus salicifolia 'Pendula'       1
      Chamaecyparis lawsoniana 'unid    1
      Platanus x hispanica Tremonia     1
      Vacant Tree Pit (planned: Gymnocladus dioicus) 1
      Vacant Tree Pit (planned: Liquidambar styraciflua) 1
      Name: Scientific Name, Length: 560, dtype: int64
```

Scientific name is qualitative **nominal**

```
[138]: names["Common Name"].value_counts()
```

```
[138]: Cherry                          12
      Rowan                           10
      Magnolia                        10
      Vacant Tree Pit (planned: )     10
      Apple - Crab                     9
      ..
      Pittosporum                      1
      Birch - Purple                    1
```

```

Maple - Column Norway      1
Maple - Crimson King Norway 1
Castlewellan gold         1
Name: Common Name, Length: 431, dtype: int64

```

Common name is qualitative **nominal**

3.3.b (Names) Descriptive Stats for Numeric Type Columns There are no numeric columns.

0.3 Task 4: Identify Missing Values

0.3.1 4.1 Missing Values for the Trees Dataset

```

[139]: # Percentage of null values
trees.isnull().mean()*100

```

```

[139]: Identifier      0.000000
Number Of Trees      0.093841
Site Name            0.000000
Contract Area        0.000000
Scientific Name       0.000000
Inspection Date      1.710459
Inspection Due Date  1.710459
Height In Metres     1.868282
Spread In Metres     1.868282
Diameter In Centimetres At Breast Height 1.872547
Ward Code            0.963999
Ward Name            0.963999
Easting              0.000000
Northing             0.000000
Longitude            0.238867
Latitude             0.238867
Location             0.238867
dtype: float64

```

```

[140]: # Number of null values
trees.isnull().sum()

```

```

[140]: Identifier      0
Number Of Trees      22
Site Name            0
Contract Area        0
Scientific Name       0
Inspection Date      401
Inspection Due Date  401
Height In Metres     438
Spread In Metres     438

```

Diameter In Centimetres At Breast Height	439
Ward Code	226
Ward Name	226
Easting	0
Northing	0
Longitude	56
Latitude	56
Location	56
dtype:	int64

```
[141]: # Percentage of zero values
trees.isin([0]).mean()*100
```

[141]: Identifier	0.000000
Number Of Trees	0.396690
Site Name	0.000000
Contract Area	0.000000
Scientific Name	0.000000
Inspection Date	0.000000
Inspection Due Date	0.000000
Height In Metres	0.733663
Spread In Metres	1.181539
Diameter In Centimetres At Breast Height	1.164477
Ward Code	0.000000
Ward Name	0.000000
Easting	0.238867
Northing	0.238867
Longitude	0.000000
Latitude	0.000000
Location	0.000000
dtype:	float64

```
[142]: # Number of zero values
trees.isin([0]).sum()
```

[142]: Identifier	0
Number Of Trees	93
Site Name	0
Contract Area	0
Scientific Name	0
Inspection Date	0
Inspection Due Date	0
Height In Metres	172
Spread In Metres	277
Diameter In Centimetres At Breast Height	273
Ward Code	0
Ward Name	0

Easting	56
Northing	56
Longitude	0
Latitude	0
Location	0
dtype: int64	

```
[143]: # Percentage of null and zero values
(trees.isnull().sum() + trees.isin([0]).sum())/trees.shape[0]
```

[143]: Identifier	0.000000
Number Of Trees	0.004905
Site Name	0.000000
Contract Area	0.000000
Scientific Name	0.000000
Inspection Date	0.017105
Inspection Due Date	0.017105
Height In Metres	0.026019
Spread In Metres	0.030498
Diameter In Centimetres At Breast Height	0.030370
Ward Code	0.009640
Ward Name	0.009640
Easting	0.002389
Northing	0.002389
Longitude	0.002389
Latitude	0.002389
Location	0.002389
dtype: float64	

```
[144]: # Number of null and zero values
(trees.isnull().sum() + trees.isin([0]).sum())
```

[144]: Identifier	0
Number Of Trees	115
Site Name	0
Contract Area	0
Scientific Name	0
Inspection Date	401
Inspection Due Date	401
Height In Metres	610
Spread In Metres	715
Diameter In Centimetres At Breast Height	712
Ward Code	226
Ward Name	226
Easting	56
Northing	56
Longitude	56

Latitude	56
Location	56
dtype: int64	

0.3.2 4.2 Missing Values for the Environmental Dataset

```
[145]: # Percentage of null values
# TODO
environmental.isnull().mean()*100
```

```
[145]: Identifier          0.000000
Maturity                1.746744
Physiological Condition  2.015802
Tree Set To Be Removed  0.000000
Removal Reason          99.641256
Capital Asset Value For Amenity Trees  1.849242
Carbon Storage In Kilograms  12.214392
Gross Carbon Sequestration Per Year In Kilograms  12.214392
Pollution Removal Per Year In Grams  12.214392
dtype: float64
```

```
[146]: # Number of null values
# TODO
environmental.isnull().sum()
```

```
[146]: Identifier          0
Maturity                409
Physiological Condition  472
Tree Set To Be Removed  0
Removal Reason          23331
Capital Asset Value For Amenity Trees  433
Carbon Storage In Kilograms  2860
Gross Carbon Sequestration Per Year In Kilograms  2860
Pollution Removal Per Year In Grams  2860
dtype: int64
```

```
[147]: # Percentage of zero values
# TODO
environmental.isin([0]).mean()*100
```

```
[147]: Identifier          0.000000
Maturity                0.000000
Physiological Condition  0.000000
Tree Set To Be Removed  0.000000
Removal Reason          0.000000
Capital Asset Value For Amenity Trees  1.183002
Carbon Storage In Kilograms  0.000000
```

Gross Carbon Sequestration Per Year In Kilograms	0.025625
Pollution Removal Per Year In Grams	0.000000
dtype: float64	

```
[148]: # Number of zero values
# TODO
environmental.isin([0]).sum()
```

[148]: Identifier	0
Maturity	0
Physiological Condition	0
Tree Set To Be Removed	0
Removal Reason	0
Capital Asset Value For Amenity Trees	277
Carbon Storage In Kilograms	0
Gross Carbon Sequestration Per Year In Kilograms	6
Pollution Removal Per Year In Grams	0
dtype: int64	

```
[149]: # Percentage of null and zero values
# TODO
(environmental.isnull().sum() + environmental.isin([0]).sum())/environmental.
↪shape[0]
```

[149]: Identifier	0.000000
Maturity	0.017467
Physiological Condition	0.020158
Tree Set To Be Removed	0.000000
Removal Reason	0.996413
Capital Asset Value For Amenity Trees	0.030322
Carbon Storage In Kilograms	0.122144
Gross Carbon Sequestration Per Year In Kilograms	0.122400
Pollution Removal Per Year In Grams	0.122144
dtype: float64	

```
[150]: # Number of null and zero values
# TODO
environmental.isnull().sum() + environmental.isin([0]).sum()
```

[150]: Identifier	0
Maturity	409
Physiological Condition	472
Tree Set To Be Removed	0
Removal Reason	23331
Capital Asset Value For Amenity Trees	710
Carbon Storage In Kilograms	2860
Gross Carbon Sequestration Per Year In Kilograms	2866

Pollution Removal Per Year In Grams
dtype: int64

2860

0.3.3 4.3 Missing Values for the Common Names Dataset

```
[151]: #Percentage of null values  
names.isnull().mean()*100
```

```
[151]: Scientific Name    0.000000  
Common Name          4.074703  
dtype: float64
```

```
[152]: #Number of null values  
names.isnull().sum()
```

```
[152]: Scientific Name    0  
Common Name           24  
dtype: int64
```

```
[153]: #Percentage of zero values  
names.isin([0]).mean()*100
```

```
[153]: Scientific Name    0.0  
Common Name           0.0  
dtype: float64
```

```
[154]: #Number of zero values  
names.isin([0]).sum()
```

```
[154]: Scientific Name    0  
Common Name           0  
dtype: int64
```

```
[155]: #Percentage of null and zero values  
(names.isnull().sum() + names.isin([0]).sum())/names.shape[0]
```

```
[155]: Scientific Name    0.000000  
Common Name          0.040747  
dtype: float64
```

```
[156]: #Number of null and zero values  
names.isnull().sum() + names.isin([0]).sum()
```

```
[156]: Scientific Name    0  
Common Name           24  
dtype: int64
```

0.3.4 4.4 Observations

Trees data:

Columns that have the greatest null and zero values:

- Inspection Date- 401
- Inspection Due Date- 401
- Height In Metres- 610
- Spread In Metres- 715
- Diameter In Centimetres At Breast Height- 712

Columns that have a moderate null and zero value (less than 401)

- Number of Trees- 115
- Ward Code- 226
- Ward Name- 226
- Easting- 56
- Northing- 56
- Longitude- 56
- Latitude- 56
- Location- 56

Environment data:

Columns that have the greatest null and zero values:

- Removal Reason- 23331
- Carbon Storage In Kilograms- 2860
- Gross Carbon Sequestration Per Year In Kilograms- 2866
- Pollution Removal Per Year In Grams- 2860

Columns that have a moderate null and zero value (less than 2860)

- Maturity- 409
- Physiological Condition- 472
- Capital Asset Value For Amenity Trees- 710

Common names data:

Columns that have the greatest null and zero values: - Common name - 24

How can this affect the councils initiative and projects?

Public Tree Data Due to the missing data, this could affect the ammount of information displayed on the public tree data section on the council website.

Tree walks brochure Since this one is focused on a guided walk, the missing data regarding the common name, location, latitude and logitude of the trees could cause some issues since the users may not be able to find certain trees they're looking for.

Environment Report Due to the missing data regarding the removal reason, carbon storage, pollution removal and physiological condition, this could cause bias in the environment report which could lead to inaccuracies and a issue of trust in the council.

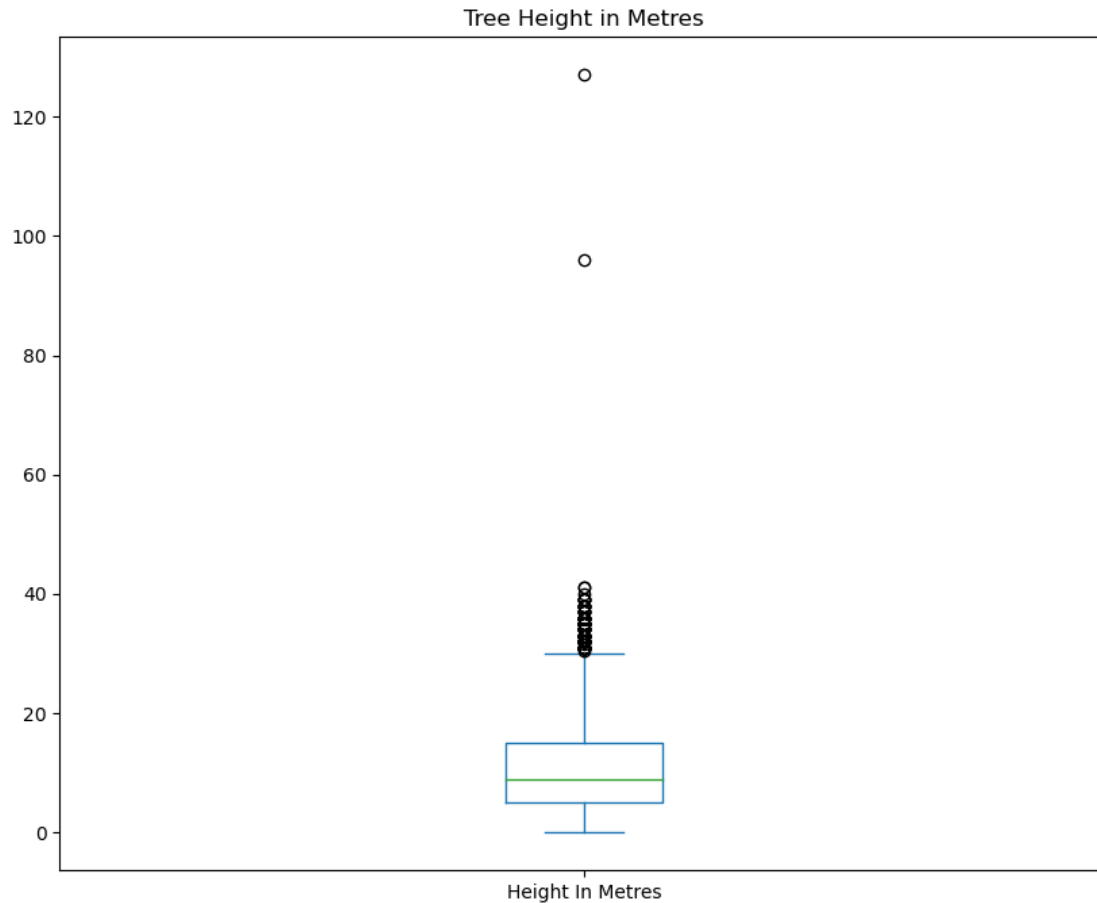
0.4 Task 5: Identify Outliers in the Trees Dimensions

0.4.1 5.1 Outliers for Height

```
[157]: #Used a Boxplot to find the outliers

trees["Height In Metres"].plot(kind='box', title='Tree Height in Metres',
    ↪figsize=(10,8))
```

```
[157]: <AxesSubplot:title={'center':'Tree Height in Metres'}>
```



```
[158]: # Selected the crazy outlier rows
mask=trees["Height In Metres"]>90
trees[mask]
```

```
[158]:
```

	Identifier	Number Of Trees	Site Name	Contract Area	\
1356	00001547	1.0	Beaumont Walk Estate	Housing	
1863	00013862	1.0	Maitland Park Estate 1	Housing	

	Scientific Name	Inspection Date	Inspection Due Date	\
1356	Robinia pseudoacacia	2017-05-23	2020/2021	
1863	Prunus avium	2017-05-16	2020/2021	

	Height In Metres	Spread In Metres	\
1356	96.0	10.0	
1863	127.0	9.0	

	Diameter In Centimetres At Breast Height	Ward Code	Ward Name	\
1356	63.0	E05000136	Haverstock	

1863

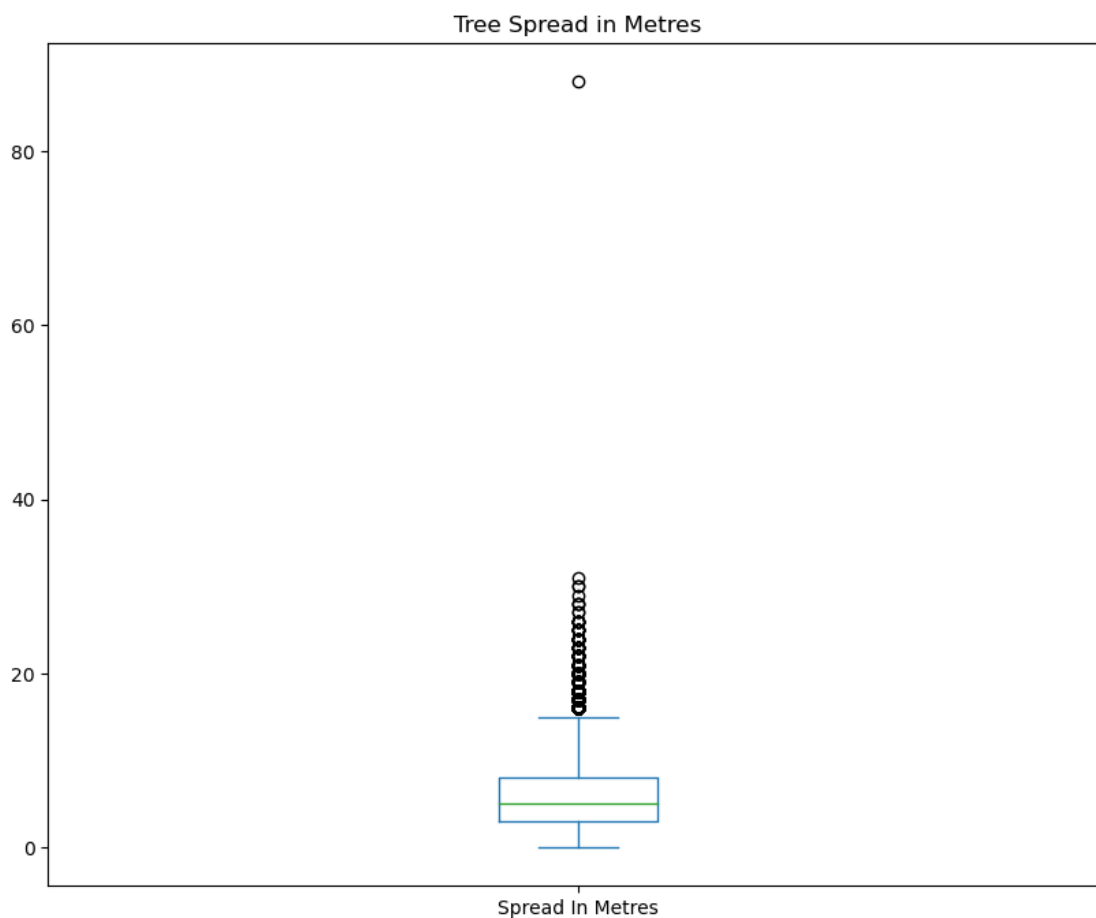
34.0 E05000136 Haverstock

	Easting	Northing	Longitude	Latitude	Location
1356	527847	184391	-0.157739	51.543941	(51.543941, -0.157739)
1863	527987	184901	-0.155534	51.548497	(51.548497, -0.155534)

0.4.2 5.2 Outliers for Spread

```
[159]: # Used a boxplot to find the outliers
trees["Spread In Metres"].plot(kind='box', title='Tree Spread in Metres',
    figsize=(10,8))
```

```
[159]: <AxesSubplot:title={'center': 'Tree Spread in Metres'}>
```



```
[160]: # Selected the crazy outlier rows (if any)
mask=trees["Spread In Metres"]>80
trees[mask]
```

```

[160]: Identifier Number Of Trees Site Name Contract Area \
18567 00045515 1.0 Broadfield Estate 1 Housing

Scientific Name Inspection Date Inspection Due Date Height In Metres \
18567 Quercus robur 2018-04-26 2021/2022 8.0

Spread In Metres Diameter In Centimetres At Breast Height Ward Code \
18567 88.0 17.0 E05000144

Ward Name Easting Northing Longitude Latitude \
18567 Swiss Cottage 525993 184693 -0.184348 51.547074

Location
18567 (51.547074, -0.184348)

```

0.4.3 5.3 Outliers for Diameter

```

[161]: # Used a boxplot to find the outliers
trees["Diameter In Centimetres At Breast Height"].plot(kind='box',
↳title='Diameters in Centimetres at Breast height', figsize=(10,8))

[161]: <AxesSubplot:title={'center': 'Diameters in Centimetres at Breast height'}>

```




```
[162]: # Selected the outlier rows (if any)
mask=trees["Diameter In Centimetres At Breast Height"]>170
trees[mask]
```

```
[162]:
```

	Identifier	Number Of Trees	Site Name \
1157	00004100	1.0	LONGFORD STREET, CLARENCE GDNS (LS)
5229	00002735	1.0	BRUNSWICK SQUARE, GARDENS (LS)
5859	00017570	1.0	RUSSELL SQUARE, GARDENS (LS)
8179	00012846	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)
8737	00019491	1.0	PANCRAS ROAD, ST. PANCRAS GDNS (LS)
8870	00017656	1.0	RUSSELL SQUARE, GARDENS (LS)
11665	00020242	1.0	TAVISTOCK SQUARE, GARDENS (LS)
11860	00012885	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)
14178	00012891	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)
15853	00012939	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)
16200	00012833	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)
17816	00021404	1.0	WATERLOW PARK (LS)
18081	00012870	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)

18930	00018775	1.0	ST. GEORGE'S GARDENS, HEATHCOTE ST (LS)
20426	00012873	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)
22781	00012923	1.0	LINCOLN'S INN FIELDS, GARDENS (LS)

	Contract Area	Scientific Name	Inspection Date	Inspection Due Date	\
1157	Parks	Platanus x hispanica	2018-03-29	2020/2021	
5229	Parks	Platanus x hispanica	2018-03-23	2020/2021	
5859	Parks	Platanus x hispanica	2018-03-14	2020/2021	
8179	Parks	Platanus x hispanica	2018-04-24	2021/2022	
8737	Parks	Platanus x hispanica	2017-03-14	2019/2020	
8870	Parks	Platanus x hispanica	2018-03-20	2020/2021	
11665	Parks	Platanus x hispanica	2018-06-04	2021/2022	
11860	Parks	Platanus x hispanica	2018-04-20	2021/2022	
14178	Parks	Platanus x hispanica	2018-04-20	2021/2022	
15853	Parks	Platanus x hispanica	2018-04-19	2021/2022	
16200	Parks	Platanus x hispanica	2018-04-25	2021/2022	
17816	Parks	Platanus x hispanica	2019-06-19	2022/2023	
18081	Parks	Platanus x hispanica	2018-04-20	2021/2022	
18930	Parks	Platanus x hispanica	2018-03-20	2020/2021	
20426	Parks	Platanus x hispanica	2018-04-20	2021/2022	
22781	Parks	Platanus x hispanica	2018-04-19	2021/2022	

	Height In Metres	Spread In Metres	\
1157	21.0	20.0	
5229	28.0	30.0	
5859	38.0	16.0	
8179	28.0	25.0	
8737	24.0	25.0	
8870	36.0	21.0	
11665	28.0	24.0	
11860	24.0	24.0	
14178	30.0	20.0	
15853	23.0	20.0	
16200	29.0	22.0	
17816	30.0	25.0	
18081	28.0	18.0	
18930	35.0	24.0	
20426	29.0	26.0	
22781	27.0	22.0	

	Diameter In Centimetres At Breast Height	Ward Code	\
1157	228.0	E05000142	
5229	197.0	E05000141	
5859	191.0	E05000129	
8179	200.0	E05000138	
8737	200.0	E05000143	
8870	177.0	E05000129	

11665	187.0	E05000129
11860	210.0	E05000138
14178	206.0	E05000138
15853	209.0	E05000138
16200	184.0	E05000138
17816	173.0	E05000137
18081	192.0	E05000138
18930	194.0	E05000141
20426	197.0	E05000138
22781	185.0	E05000138

	Ward Name	Easting	Northing	Longitude	Latitude \
1157	Regent's Park	528931	182624	-0.142766	51.527816
5229	King's Cross	530387	182261	-0.121914	51.524218
5859	Bloomsbury	530025	181952	-0.127255	51.521529
8179	Holborn and Covent Garden	530834	181409	-0.115802	51.516467
8737	St Pancras and Somers Town	529789	183478	-0.130079	51.535299
8870	Bloomsbury	530122	181955	-0.125851	51.521536
11665	Bloomsbury	529873	182306	-0.129313	51.524743
11860	Holborn and Covent Garden	530820	181358	-0.116011	51.516010
14178	Holborn and Covent Garden	530783	181341	-0.116553	51.515861
15853	Holborn and Covent Garden	530705	181373	-0.117672	51.516165
16200	Holborn and Covent Garden	530816	181440	-0.116050	51.516750
17816	Highgate	528815	187160	-0.142769	51.568608
18081	Holborn and Covent Garden	530873	181365	-0.115257	51.516056
18930	King's Cross	530485	182497	-0.120423	51.526317
20426	Holborn and Covent Garden	530853	181378	-0.115534	51.516179
22781	Holborn and Covent Garden	530769	181399	-0.116734	51.516390

	Location
1157	(51.527816, -0.142766)
5229	(51.524218, -0.121914)
5859	(51.521529, -0.127255)
8179	(51.516467, -0.115802)
8737	(51.535299, -0.130079)
8870	(51.521536, -0.125851)
11665	(51.524743, -0.129313)
11860	(51.51601, -0.116011)
14178	(51.515861, -0.116553)
15853	(51.516165, -0.117672)
16200	(51.51675, -0.11605)
17816	(51.568608, -0.142769)
18081	(51.516056, -0.115257)
18930	(51.526317, -0.120423)
20426	(51.516179, -0.115534)
22781	(51.51639, -0.116734)

0.4.4 5.4 Observations

The outliers in all three columns (Height in Metres, Spread in Metres and Diameter In Centimetres At Breast Height) were easy to identify. This was because I noticed that most of the data points above the maximum were tightly connected to each other, so the data points that were further away and segregated must have been the outlier since it wasn't linked to the previous data points.

The column that had the most outliers was the Diameter in centimetres at breast height. However, the outliers for the Diameter in cm at breast height is difficult to justify whether or not it is an incorrect input since all the outliers seem to be linked closely. Hence, I chose 170 as my threshold value for the outliers in this case.

The Height and Spread columns seemed to show the fewest outliers.

The Height columns outlier must have been inputted correctly, since there are no trees above the height of 100 metres in Camden. This matter could be possible for the Spread column too since the outlier is segregated from the rest of the circles, which could indicate an incorrect value too.

0.5 Task 6: Identify Duplicates in the Trees Dataset

0.5.1 6.1 Find Duplicate Rows

In our dataset the Identifier column should be unique. Find out if it is! We've already used a function that can count how many times each value in a column exists. Use it to see if we have duplicates in the trees DataFrame.

```
[163]: # Found out if we have any duplicates
trees["Identifier"].duplicated().sum()
```

```
[163]: 6
```

```
[164]: identifier_counts = trees["Identifier"].value_counts()
identifier_counts[identifier_counts>1]
```

```
[164]: 00000999    2
00060087    2
00022744    2
00032549    2
00022674    2
00060088    2
Name: Identifier, dtype: int64
```

Now see if you can select the rows from trees DataFrame that are duplicates. You will need to use the output from the cell above and use it to filter the trees dataframe.

```
[165]: # Selected the rows that are duplicated
trees[trees["Identifier"].duplicated()]
```

```
[165]:
```

	Identifier	Number Of Trees	Site Name	Contract Area	\
6111	00032549	1.0	NARCISSUS ROAD	Highways	
9186	00022744	1.0	YORK WAY	Highways	

10972	00060088	1.0	FREDERICK STREET	Highways
13098	00000999	1.0	ALMA STREET	Highways
13628	00022674	1.0	WOODSOME ROAD	Highways
15653	00060087	1.0	ARGYLE SQUARE	Highways

	Scientific Name	Inspection Date	\
6111	Betula albosinensis Fasc.	2018-09-19	
9186	Ailanthus altissima	2019-10-30	
10972	Vacant Tree Pit (planned: Acer campestre eco s...	2019-11-09	
13098	Sorbus hupehensis	2017-07-25	
13628	Sorbus	2017-10-07	
15653	Vacant Tree Pit (planned: Acer campestre eco s...	2019-11-09	

	Inspection Due Date	Height In Metres	Spread In Metres	\
6111	2021/2022	9.0	6.0	
9186	2022/2023	7.5	3.0	
10972	2022/2023	NaN	NaN	
13098	2020/2021	5.0	4.0	
13628	2020/2021	7.0	6.0	
15653	2022/2023	NaN	NaN	

	Diameter In Centimetres At Breast Height	Ward Code	Ward Name	\
6111	19.0	E05000145	West Hampstead	
9186	18.0	E05000131	Cantelowes	
10972	NaN	E05000141	King's Cross	
13098	18.0	E05000139	Kentish Town	
13628	28.0	E05000137	Highgate	
15653	NaN	E05000141	King's Cross	

	Easting	Northing	Longitude	Latitude	Location
6111	525185	185127	-0.195840	51.551159	(51.551159, -0.19584)
9186	529983	184724	-0.126837	51.546447	(51.546447, -0.126837)
10972	530770	182696	-0.116241	51.528046	(51.528046, -0.116241)
13098	528834	184856	-0.143337	51.547904	(51.547904, -0.143337)
13628	528515	186109	-0.147481	51.559237	(51.559237, -0.147481)
15653	530342	182839	-0.122352	51.529428	(51.529428, -0.122352)

0.5.2 6.2 Observations

There were 6 duplicates in the Identifier column in the Tree data set. Luckily, it was only duplicated once. The identifiers that had the duplicates were: 00000999

00060087
00022744
00032549
00022674
00060088

These duplicates would have to be removed from the data set before they are used in the council's

projects or during analysis since this could cause inaccuracies in the data and subsequent analysis

0.6 Task 7: Identify Geolocation Issues

The geographic coordinates (Easting and Northing) can be used to plot the trees on a map. We can use this approach to see if there are any unusual tree locations!

0.6.1 7.1 Remove Trees with Missing Geo-coordinates

```
[166]: # Made a copy of the trees
geotrees = trees.copy()
```

```
[167]: # Removed null Eastings

mask = geotrees["Easting"].notnull()
geotrees[mask]
```

```
[167]:
```

	Identifier	Number Of Trees	Site Name \
0	00060053	1.0	Russell Nurseries Estate
1	00057855	1.0	BRECKNOCK JMI (E)
2	00059953	1.0	Estate 51 Ravenshaw Street
3	00059915	1.0	ROSARY RC JMI (E)
4	00010762	1.0	Holly Lodge Estate
...
23439	00057455	1.0	KILBURN GRANGE, MESSINA AVE (LS)
23440	00015494	1.0	OSSULSTON STREET
23441	00001693	1.0	BELMONT STREET
23442	00020342	1.0	Templar House Estate
23443	00013369	1.0	LYNDHURST ROAD

	Contract Area	Scientific Name	Inspection Date \
0	Housing	Vacant Tree Pit	NaT
1	Education	Vacant Tree Pit	2019-07-17
2	Housing	Ficus carica	NaT
3	Education	Betula jacquemontii	NaT
4	Housing	Ilex x altaclarensis	2017-06-14
...
23439	Parks	Sorbus aria	2017-03-28
23440	Highways	Sorbus aria 'Majestica'	2019-10-28
23441	Highways	Platanus x hispanica	2017-07-31
23442	Housing	Tilia europaea	2018-11-05
23443	Highways	Tilia europaea	2018-06-08

	Inspection Due Date	Height In Metres	Spread In Metres \
0	NaN	NaN	NaN
1	2022/2023	NaN	NaN
2	NaN	5.0	4.0
3	NaN	4.0	1.0

4	2020/2021	14.0	6.0
...
23439	2019/2020	2.0	1.0
23440	2022/2023	12.0	6.0
23441	2020/2021	18.0	10.0
23442	2021/2022	20.0	8.0
23443	2021/2022	15.0	6.0

	Diameter In Centimetres At Breast Height	Ward Code \
0	NaN	E05000135
1	NaN	E05000131
2	10.0	NaN
3	6.0	E05000135
4	26.0	E05000137
...
23439	6.0	E05000140
23440	48.0	E05000143
23441	57.0	E05000136
23442	40.0	E05000132
23443	77.0	E05000135

	Ward Name	Easting	Northing	Longitude	Latitude \
0	Hampstead Town	527305	185240	-0.165240	51.551693
1	Cantelowes	529923	184782	-0.127681	51.546984
2	NaN	0	0	NaN	NaN
3	Hampstead Town	527249	185261	-0.166051	51.551901
4	Highgate	528414	186770	-0.148704	51.565198
...
23439	Kilburn	525130	184418	-0.196884	51.544796
23440	St Pancras and Somers Town	529758	183095	-0.130667	51.531863
23441	Haverstock	528302	184457	-0.151163	51.544432
23442	Fortune Green	524615	184714	-0.204206	51.547573
23443	Hampstead Town	526738	185304	-0.173397	51.552397

	Location
0	(51.551693, -0.16524)
1	(51.546984, -0.127681)
2	NaN
3	(51.551901, -0.166051)
4	(51.565198, -0.148704)
...	...
23439	(51.544796, -0.196884)
23440	(51.531863, -0.130667)
23441	(51.544432, -0.151163)
23442	(51.547573, -0.204206)
23443	(51.552397, -0.173397)

[23444 rows x 17 columns]

```
[168]: # Removed 0 Eastings.
mask = geotrees["Easting"] !=0
geotrees[mask]
```

```
[168]: Identifier Number Of Trees Site Name \
0 00060053 1.0 Russell Nurseries Estate
1 00057855 1.0 BRECKNOCK JMI (E)
3 00059915 1.0 ROSARY RC JMI (E)
4 00010762 1.0 Holly Lodge Estate
5 00007523 1.0 Westcroft Estate 1
...
23439 00057455 1.0 KILBURN GRANGE, MESSINA AVE (LS)
23440 00015494 1.0 OSSULSTON STREET
23441 00001693 1.0 BELMONT STREET
23442 00020342 1.0 Templar House Estate
23443 00013369 1.0 LYNDHURST ROAD
```

```
Contract Area Scientific Name Inspection Date \
0 Housing Vacant Tree Pit NaT
1 Education Vacant Tree Pit 2019-07-17
3 Education Betula jacquemontii NaT
4 Housing Ilex x altaclarensis 2017-06-14
5 Housing Betula pendula 2018-08-06
...
23439 Parks Sorbus aria 2017-03-28
23440 Highways Sorbus aria 'Majestica' 2019-10-28
23441 Highways Platanus x hispanica 2017-07-31
23442 Housing Tilia europaea 2018-11-05
23443 Highways Tilia europaea 2018-06-08
```

```
Inspection Due Date Height In Metres Spread In Metres \
0 NaN NaN NaN
1 2022/2023 NaN NaN
3 NaN 4.0 1.0
4 2020/2021 14.0 6.0
5 2021/2022 9.0 7.0
...
23439 2019/2020 2.0 1.0
23440 2022/2023 12.0 6.0
23441 2020/2021 18.0 10.0
23442 2021/2022 20.0 8.0
23443 2021/2022 15.0 6.0
```

```
Diameter In Centimetres At Breast Height Ward Code \
0 NaN E05000135
```



```

1          NaN  E05000131
3          6.0  E05000135
4         26.0  E05000137
5         29.0          NaN
...
23439      6.0  E05000140
23440     48.0  E05000143
23441     57.0  E05000136
23442     40.0  E05000132
23443     77.0  E05000135

```

	Ward Name	Easting	Northing	Longitude	Latitude \
0	Hampstead Town	527305	185240	-0.165240	51.551693
1	Cantelowes	529923	184782	-0.127681	51.546984
3	Hampstead Town	527249	185261	-0.166051	51.551901
4	Highgate	528414	186770	-0.148704	51.565198
5	NaN	524253	185982	-0.208975	51.559049
...
23439	Kilburn	525130	184418	-0.196884	51.544796
23440	St Pancras and Somers Town	529758	183095	-0.130667	51.531863
23441	Haverstock	528302	184457	-0.151163	51.544432
23442	Fortune Green	524615	184714	-0.204206	51.547573
23443	Hampstead Town	526738	185304	-0.173397	51.552397

	Location
0	(51.551693, -0.16524)
1	(51.546984, -0.127681)
3	(51.551901, -0.166051)
4	(51.565198, -0.148704)
5	(51.559049, -0.208975)
...	...
23439	(51.544796, -0.196884)
23440	(51.531863, -0.130667)
23441	(51.544432, -0.151163)
23442	(51.547573, -0.204206)
23443	(51.552397, -0.173397)

[23388 rows x 17 columns]

```

[169]: # Removed null Northings
mask = geotrees["Northing"].notnull()
geotrees[mask]

```

	Identifier	Number Of Trees	Site Name \
0	00060053	1.0	Russell Nurseries Estate
1	00057855	1.0	BRECKNOCK JMI (E)
2	00059953	1.0	Estate 51 Ravenshaw Street

3	00059915	1.0	ROSARY RC JMI (E)
4	00010762	1.0	Holly Lodge Estate
...
23439	00057455	1.0	KILBURN GRANGE, MESSINA AVE (LS)
23440	00015494	1.0	OSSULSTON STREET
23441	00001693	1.0	BELMONT STREET
23442	00020342	1.0	Templar House Estate
23443	00013369	1.0	LYNDHURST ROAD

	Contract Area	Scientific Name	Inspection Date \
0	Housing	Vacant Tree Pit	NaT
1	Education	Vacant Tree Pit	2019-07-17
2	Housing	Ficus carica	NaT
3	Education	Betula jacquemontii	NaT
4	Housing	Ilex x altaclarensis	2017-06-14
...
23439	Parks	Sorbus aria	2017-03-28
23440	Highways	Sorbus aria 'Majestica'	2019-10-28
23441	Highways	Platanus x hispanica	2017-07-31
23442	Housing	Tilia europaea	2018-11-05
23443	Highways	Tilia europaea	2018-06-08

	Inspection Due Date	Height In Metres	Spread In Metres \
0	NaT	NaT	NaT
1	2022/2023	NaT	NaT
2	NaT	5.0	4.0
3	NaT	4.0	1.0
4	2020/2021	14.0	6.0
...
23439	2019/2020	2.0	1.0
23440	2022/2023	12.0	6.0
23441	2020/2021	18.0	10.0
23442	2021/2022	20.0	8.0
23443	2021/2022	15.0	6.0

	Diameter In Centimetres At Breast Height	Ward Code \
0	NaT	E05000135
1	NaT	E05000131
2	10.0	NaT
3	6.0	E05000135
4	26.0	E05000137
...
23439	6.0	E05000140
23440	48.0	E05000143
23441	57.0	E05000136
23442	40.0	E05000132
23443	77.0	E05000135

	Ward Name	Easting	Northing	Longitude	Latitude \
0	Hampstead Town	527305	185240	-0.165240	51.551693
1	Cantelowes	529923	184782	-0.127681	51.546984
2	NaN	0	0	NaN	NaN
3	Hampstead Town	527249	185261	-0.166051	51.551901
4	Highgate	528414	186770	-0.148704	51.565198
...
23439	Kilburn	525130	184418	-0.196884	51.544796
23440	St Pancras and Somers Town	529758	183095	-0.130667	51.531863
23441	Haverstock	528302	184457	-0.151163	51.544432
23442	Fortune Green	524615	184714	-0.204206	51.547573
23443	Hampstead Town	526738	185304	-0.173397	51.552397

	Location
0	(51.551693, -0.16524)
1	(51.546984, -0.127681)
2	NaN
3	(51.551901, -0.166051)
4	(51.565198, -0.148704)
...	...
23439	(51.544796, -0.196884)
23440	(51.531863, -0.130667)
23441	(51.544432, -0.151163)
23442	(51.547573, -0.204206)
23443	(51.552397, -0.173397)

[23444 rows x 17 columns]

```
[170]: # Removed 0 Northings.
mask = geotrees["Northing"] !=0
geotrees[mask]
```

[170]:	Identifier	Number Of Trees	Site Name \
0	00060053	1.0	Russell Nurseries Estate
1	00057855	1.0	BRECKNOCK JMI (E)
3	00059915	1.0	ROSARY RC JMI (E)
4	00010762	1.0	Holly Lodge Estate
5	00007523	1.0	Westcroft Estate 1
...
23439	00057455	1.0	KILBURN GRANGE, MESSINA AVE (LS)
23440	00015494	1.0	OSSULSTON STREET
23441	00001693	1.0	BELMONT STREET
23442	00020342	1.0	Templar House Estate
23443	00013369	1.0	LYNDHURST ROAD

Contract Area	Scientific Name	Inspection Date \
---------------	-----------------	-------------------

0	Housing	Vacant Tree Pit	NaT
1	Education	Vacant Tree Pit	2019-07-17
3	Education	Betula jacquemontii	NaT
4	Housing	Ilex x altaclarensis	2017-06-14
5	Housing	Betula pendula	2018-08-06
...
23439	Parks	Sorbus aria	2017-03-28
23440	Highways	Sorbus aria 'Majestica'	2019-10-28
23441	Highways	Platanus x hispanica	2017-07-31
23442	Housing	Tilia europaea	2018-11-05
23443	Highways	Tilia europaea	2018-06-08

	Inspection Due Date	Height In Metres	Spread In Metres \
0	NaN	NaN	NaN
1	2022/2023	NaN	NaN
3	NaN	4.0	1.0
4	2020/2021	14.0	6.0
5	2021/2022	9.0	7.0
...
23439	2019/2020	2.0	1.0
23440	2022/2023	12.0	6.0
23441	2020/2021	18.0	10.0
23442	2021/2022	20.0	8.0
23443	2021/2022	15.0	6.0

	Diameter In Centimetres At Breast Height	Ward Code \
0	NaN	E05000135
1	NaN	E05000131
3	6.0	E05000135
4	26.0	E05000137
5	29.0	NaN
...
23439	6.0	E05000140
23440	48.0	E05000143
23441	57.0	E05000136
23442	40.0	E05000132
23443	77.0	E05000135

	Ward Name	Easting	Northing	Longitude	Latitude \
0	Hampstead Town	527305	185240	-0.165240	51.551693
1	Cantelowes	529923	184782	-0.127681	51.546984
3	Hampstead Town	527249	185261	-0.166051	51.551901
4	Highgate	528414	186770	-0.148704	51.565198
5	NaN	524253	185982	-0.208975	51.559049
...
23439	Kilburn	525130	184418	-0.196884	51.544796
23440	St Pancras and Somers Town	529758	183095	-0.130667	51.531863

23441	Haverstock	528302	184457	-0.151163	51.544432
23442	Fortune Green	524615	184714	-0.204206	51.547573
23443	Hampstead Town	526738	185304	-0.173397	51.552397

	Location
0	(51.551693, -0.16524)
1	(51.546984, -0.127681)
3	(51.551901, -0.166051)
4	(51.565198, -0.148704)
5	(51.559049, -0.208975)
...	...
23439	(51.544796, -0.196884)
23440	(51.531863, -0.130667)
23441	(51.544432, -0.151163)
23442	(51.547573, -0.204206)
23443	(51.552397, -0.173397)

[23388 rows x 17 columns]

```
[171]: # Confirmed how many rows we have
geotrees[mask].shape
```

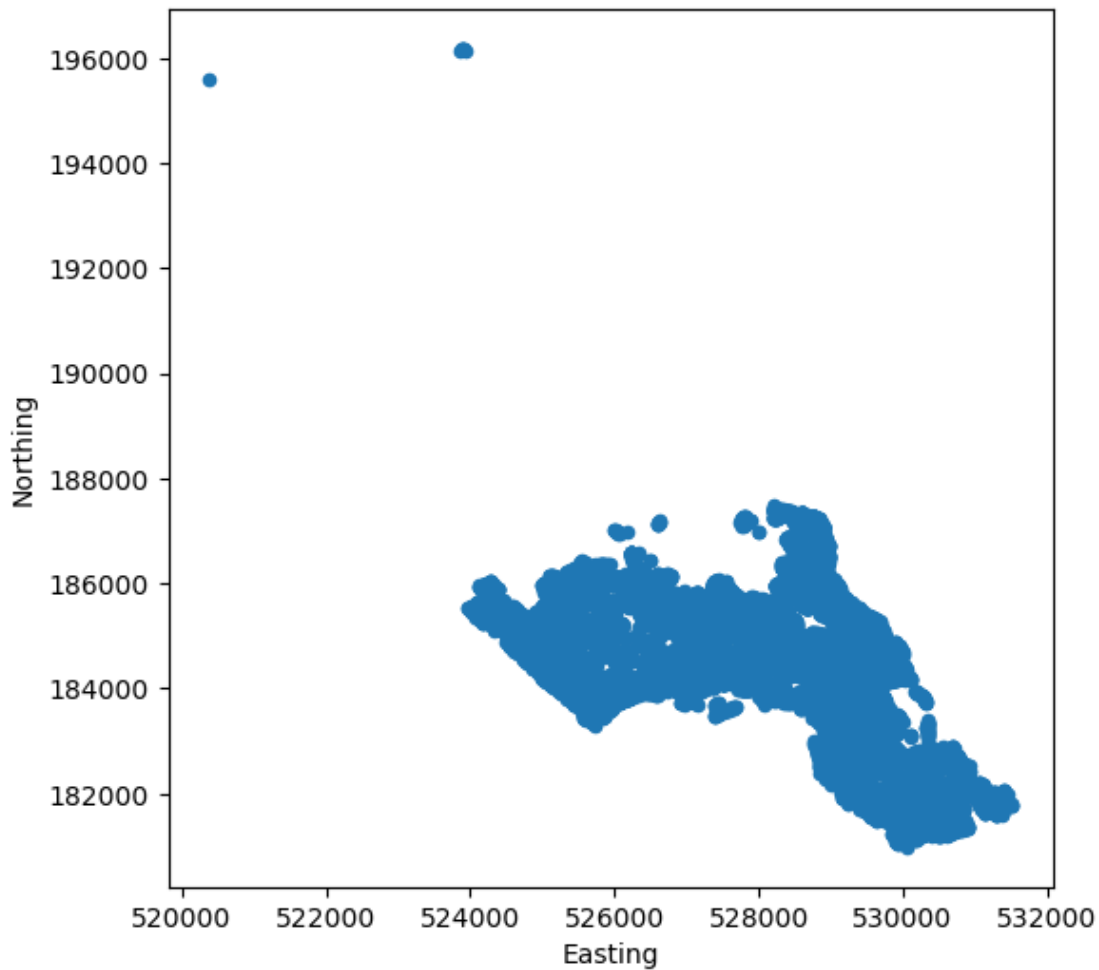
```
[171]: (23388, 17)
```

There were 56 rows in both Easting and Northing that had 0's in the values. Both had 0 rows that contained nulls. Before the null and zero values were removed, the Count for both Easting and Northing was 23,444 but **after it was removed**, there are now **23,338** rows.

0.6.2 7.2 Plot Map of Trees

```
[172]: # Ploted the trees on a map
geotrees[mask].plot.scatter(x='Easting', y="Northing", figsize=(6,6))
```

```
[172]: <AxesSubplot:xlabel='Easting', ylabel='Northing'>
```



0.6.3 7.3 Find Trees Outside Camden

From the scatter plot, you should be able to determine how to select the rows from the trees data set containing the offending trees (using the Easting and Northing values)

Select the rows containing trees outside of Camden. Use the filter technique again.

```
[173]: # Selected the outlier rows
mask = geotrees["Northing"] > 194000
geotrees[mask]
```

```
[173]:
```

	Identifier	Number Of Trees	Site Name	Contract Area	\
78	00044991	1.0	Bells Hill Estate	Housing	
298	00045000	1.0	Estate 167 Furzehill Road	Housing	
660	00044992	1.0	Bells Hill Estate	Housing	
1526	00044995	1.0	Bells Hill Estate	Housing	
5392	00044990	1.0	Bells Hill Estate	Housing	

18069	00044993	1.0	Bells Hill Estate	Housing
18078	00044601	1.0	Bells Hill Estate	Housing
19532	00044988	1.0	Bells Hill Estate	Housing

	Scientific Name	Inspection Date	Inspection Due Date	\
78	Fraxinus excelsior	2017-04-28	2020/2021	
298	Pinus sylvestris	2017-04-28	2020/2021	
660	Crataegus monogyna	2017-04-28	2020/2021	
1526	Fraxinus excelsior	2017-04-28	2020/2021	
5392	Aesculus hippocastanum	2017-04-28	2020/2021	
18069	Tilia cordata	2017-04-28	2020/2021	
18078	Tilia cordata	2017-04-28	2020/2021	
19532	Tilia cordata	2017-04-28	2020/2021	

	Height In Metres	Spread In Metres	\
78	20.0	10.0	
298	10.0	3.0	
660	4.0	4.0	
1526	18.0	12.0	
5392	22.0	12.0	
18069	23.0	14.0	
18078	21.0	12.0	
19532	21.0	12.0	

	Diameter In Centimetres At Breast	Height	Ward Code	Ward Name	Easting	\
78	49.0	NaN	NaN	NaN	523883	
298	41.0	NaN	NaN	NaN	520367	
660	7.0	NaN	NaN	NaN	523875	
1526	54.0	NaN	NaN	NaN	523936	
5392	67.0	NaN	NaN	NaN	523889	
18069	89.0	NaN	NaN	NaN	523867	
18078	56.0	NaN	NaN	NaN	523905	
19532	65.0	NaN	NaN	NaN	523909	

	Northing	Longitude	Latitude	Location
78	196179	-0.210713	51.650765	(51.650765, -0.210713)
298	195595	-0.261719	51.646283	(51.646283, -0.261719)
660	196170	-0.210830	51.650690	(51.65069, -0.21083)
1526	196127	-0.209969	51.650287	(51.650287, -0.209969)
5392	196188	-0.210618	51.650843	(51.650843, -0.210618)
18069	196159	-0.210957	51.650590	(51.65059, -0.210957)
18078	196174	-0.210394	51.650719	(51.650719, -0.210394)
19532	196169	-0.210340	51.650676	(51.650676, -0.21034)

```
[174]: # Confirmed how many rows we have
geotrees[mask].shape
```

[174]: (8, 17)

0.6.4 7.4 Observations

For the outliers in this scatter plot, there seems to be 8 trees that are outside Camden. It was easy to identify the outliers in this case as there are two distinct dots that are segregated from the rest of the “map” of camden, which shows they aren’t connected. I found the outlier by using the Northing column and by masking values that were greater than 194000 as shown on the scatter plot.

0.7 Task 8: Identify Unmatched Data

We have multiple datasets that will need to be joined together to produce the analyses required by the Camden Parks and Open Spaces team. The data will need to be joined in the following way:

- Use the **Identifier** column in the trees dataset to match to the **Identifier** column in the environmental data set (so we can bring in the environmental data for each tree)
- Use the **Scientific Name** column in the trees dataset to match to the **Scientific Name** column in the common names data set (so we can look up the **Common Name**)

There may be mismatches in the data. Of particular concern we want to check

- That every tree in the trees dataset has matching environmental data in the environmental data set
- That every environmental row in the environmental dataset has matching tree data in the tree data set
- That every scientific name in the trees dataset has a matching common name in the common names data set

0.7.1 8.1 Find Trees that Don’t have Matching Environmental Data

```
[175]: # Found trees that don't have matching environmental data

mask = ~trees["Identifier"].isin(environmental["Identifier"])

trees[mask]
```

```
[175]:
```

	Identifier	Number Of Trees	Site Name \
66	00059712	1.0	Maiden Lane Estate
125	00048578	1.0	BUCK STREET
1148	00006577	1.0	FAWLEY ROAD
1998	00007366	1.0	FORTUNE GREEN RD, OPEN SPACE (LS)
2246	00014633	1.0	Mortimer Estate
5478	00060382	1.0	SHAFTESBURY AVENUE
10637	00002874	1.0	BURGHLEY ROAD
10977	00055227	1.0	BURGHLEY ROAD
11795	00016702	1.0	RED LION SQUARE, GARDENS (LS)
11856	00054744	1.0	Carrol & Sanderson Close Estate
12056	00003694	1.0	Estate 1-161 Burnham (cons)

12936	00054558	1.0	ST. MARY'S KILBURN C OF E JMI (E)
13248	00059317	1.0	ADELAIDE ROAD NATURE AREA
16815	00055884	1.0	HONEYBOURNE ROAD
18690	00059963	1.0	Ampthill Square Estate
18958	00059246	1.0	Belsize nature reserve, Russell Nursery
19606	00005127	1.0	CUMBERLAND MARKET, OPEN SPACE (LS)
20169	00017912	1.0	SHERRIFF ROAD
20226	00047080	1.0	Ampthill Square Estate
21287	00029059	1.0	Estate 1-20 Marrick House (cons)
22470	00012126	1.0	KINGS COLLEGE ROAD
23301	00010784	1.0	Holly Lodge Estate
23315	00056485	1.0	WATERLOW PARK (LS)

	Contract Area	Scientific Name \
66	Housing	Acer saccharinum
125	Highways	Sorbus aucuparia
1148	Highways	Tilia euchlora
1998	Parks	Ilex aquifolium
2246	Housing	Tilia europaea
5478	Highways	Vacant Tree Pit
10637	Highways	Platanus x hispanica
10977	Highways	Amelanchier lamarckii
11795	Parks	Platanus x hispanica
11856	Housing	Prunus unidentified species
12056	Housing	Acer platanoides
12936	Education	Amelanchier lamarckii
13248	Parks	Stump Only
16815	Highways	Acer pseudoplatanus 'Brilliant
18690	Housing	Vacant Tree Pit (planned: Parrotia persica van...
18958	Parks	Ulmus procera
19606	Parks	Platanus x hispanica
20169	Highways	Tilia platyphyllos
20226	Housing	Malus unidentified species
21287	Housing	Sambucus nigra
22470	Highways	Fraxinus excelsior
23301	Housing	Ilex aquifolium
23315	Parks	Fraxinus excelsior

	Inspection Date	Inspection Due Date	Height In Metres	Spread In Metres \
66	2019-05-28	2022/2023	12.0	5.0
125	2017-07-19	2020/2021	6.0	2.0
1148	2018-09-28	2021/2022	15.0	6.0
1998	2017-03-21	2019/2020	9.0	6.0
2246	2019-01-29	2021/2022	16.0	12.0
5478	NaT	NaN	NaN	NaN
10637	2017-08-14	2020/2021	20.0	8.0
10977	2017-08-14	2020/2021	3.0	2.0

11795	2018-06-04	2021/2022	30.0	23.0
11856	2017-01-06	2020/2021	3.0	3.0
12056	2018-04-17	2021/2022	3.0	1.0
12936	2019-10-07	2022/2023	4.0	2.0
13248	2019-01-31	2021/2022	0.0	5.0
16815	2018-05-10	2021/2022	2.0	1.0
18690	2019-01-08	2022/2023	NaN	NaN
18958	2019-01-29	2021/2022	5.0	4.0
19606	2018-03-13	2020/2021	10.0	8.0
20169	2018-10-09	2021/2022	9.0	5.0
20226	2019-01-08	2022/2023	5.0	3.0
21287	2018-06-19	2021/2022	6.0	6.0
22470	2018-07-13	2021/2022	18.0	12.0
23301	2017-06-14	2020/2021	7.0	5.0
23315	2019-05-24	2022/2023	12.0	5.0

	Diameter In Centimetres At Breast Height	Ward Code \
66	20.0	E05000131
125	10.0	E05000130
1148	38.0	E05000145
1998	44.0	E05000132
2246	47.0	E05000140
5478	NaN	E05000138
10637	52.0	E05000139
10977	5.0	E05000139
11795	165.0	E05000138
11856	13.0	E05000137
12056	6.0	E05000128
12936	8.0	E05000140
13248	50.0	E05000128
16815	4.0	E05000145
18690	NaN	E05000143
18958	11.0	E05000134
19606	38.0	E05000142
20169	42.0	E05000145
20226	16.0	E05000143
21287	39.0	E05000140
22470	48.0	E05000128
23301	20.0	E05000137
23315	16.0	E05000137

	Ward Name	Easting	Northing	Longitude \
66	Cantelowes	529795	184142	-0.129750
125	Camden Town with Primrose Hill	528900	184020	-0.142698
1148	West Hampstead	525572	185015	-0.190313
1998	Fortune Green	525074	185541	-0.197304
2246	Kilburn	525763	183613	-0.188060

5478	Holborn and Covent Garden	530073	181247	-0.126813
10637	Kentish Town	529119	185889	-0.138863
10977	Kentish Town	528920	185694	-0.141802
11795	Holborn and Covent Garden	530572	181701	-0.119460
11856	Highgate	528661	185556	-0.145585
12056	Belsize	527015	184315	-0.169763
12936	Kilburn	525443	183919	-0.192552
13248	Belsize	527577	184297	-0.161664
16815	West Hampstead	525593	185050	-0.189987
18690	St Pancras and Somers Town	529216	183103	-0.138479
18958	Gospel Oak	527523	185233	-0.162101
19606	Regent's Park	528913	182814	-0.142959
20169	West Hampstead	525265	184543	-0.194897
20226	St Pancras and Somers Town	529279	183106	-0.137573
21287	Kilburn	525832	183583	-0.187075
22470	Belsize	526999	184416	-0.169957
23301	Highgate	528472	186812	-0.147845
23315	Highgate	528730	187264	-0.143955

	Latitude	Location
66	51.541266	(51.541266, -0.12975)
125	51.540371	(51.540371, -0.142698)
1148	51.550062	(51.550062, -0.190313)
1998	51.554901	(51.554901, -0.197304)
2246	51.537424	(51.537424, -0.18806)
5478	51.515186	(51.515186, -0.126813)
10637	51.557120	(51.55712, -0.138863)
10977	51.555408	(51.555408, -0.141802)
11795	51.519146	(51.519146, -0.11946)
11856	51.554226	(51.554226, -0.145585)
12056	51.543451	(51.543451, -0.169763)
12936	51.540245	(51.540245, -0.192552)
13248	51.543163	(51.543163, -0.161664)
16815	51.550377	(51.550377, -0.189987)
18690	51.532058	(51.532058, -0.138479)
18958	51.551587	(51.551587, -0.162101)
19606	51.529527	(51.529527, -0.142959)
20169	51.545886	(51.545886, -0.194897)
20226	51.532073	(51.532073, -0.137573)
21287	51.537139	(51.537139, -0.187075)
22470	51.544363	(51.544363, -0.169957)
23301	51.565562	(51.565562, -0.147845)
23315	51.569567	(51.569567, -0.143955)

[176]: # Confirmed how many rows we have

```
trees[mask].shape
```

[176]: (23, 17)

0.7.2 8.2 Find Environmental Data that Doesn't have Matching Tree Data

```
[184]: # Found environmental data that doesn't have matching tree data
```

```
mask = ~environmental["Identifier"].isin(trees["Identifier"])
environmental[mask]
```

[184]: Empty DataFrame

Columns: [Identifier, Maturity, Physiological Condition, Tree Set To Be Removed, Removal Reason, Capital Asset Value For Amenity Trees, Carbon Storage In Kilograms, Gross Carbon Sequestration Per Year In Kilograms, Pollution Removal Per Year In Grams]
Index: []

```
[185]: # Confirmed how many rows we have
```

```
environmental[mask].shape
```

[185]: (0, 9)

0.7.3 8.3 Find Trees that Don't have Matching Common Names Data

```
[179]: # Found trees with scientific names that don't have matching common names data
```

```
mask = ~trees["Scientific Name"].isin(names["Scientific Name"])
trees[mask]
```

```
[179]:
```

	Identifier	Number Of Trees	Site Name \
151	00051832	1.0	ARGYLE WALK
384	00053954	1.0	CHURCHILL ROAD
495	00047497	1.0	PATSHULL PLACE
611	00055434	1.0	SHARPLES HALL STREET
653	00055289	1.0	QUEEN'S CRESCENT
...
21826	00050835	1.0	INGESTRE RD
22948	00052341	1.0	NEW COMPTON STREET
23266	00048846	1.0	ASMARA ROAD
23335	00048705	1.0	GOLDINGTON STREET
23372	00031627	1.0	ST. GEORGE THE MARTYR C OF E JMI (E)

	Contract Area	Scientific Name	Inspection Date \
151	Highways	Sorbus aucuparia 'Streetwise'	2019-02-10

384	Highways	Sorbus aucuparia	'Streetwise'	2017-10-07
495	Highways	Sorbus aucuparia	'Streetwise'	2017-06-22
611	Highways	Sorbus aucuparia	'Streetwise'	2019-09-30
653	Highways	Sorbus aucuparia	'Streetwise'	2017-07-08
...
21826	Highways	Sorbus aucuparia	'Streetwise'	2017-08-18
22948	Highways	Sorbus aucuparia	'Streetwise'	2019-07-08
23266	Highways	Sorbus aucuparia	'Streetwise'	2018-08-28
23335	Highways	Sorbus aucuparia	'Streetwise'	2019-10-23
23372	Education	Cotoneaster salicifolius		2018-07-23

	Inspection Due Date	Height In Metres	Spread In Metres	\
151	2022/2023	7.0	3.0	
384	2020/2021	3.0	2.0	
495	2020/2021	5.0	3.0	
611	2022/2023	2.0	2.0	
653	2020/2021	4.0	1.0	
...
21826	2020/2021	3.0	1.0	
22948	2022/2023	4.0	3.0	
23266	2021/2022	5.0	3.0	
23335	2022/2023	6.0	2.0	
23372	2021/2022	5.0	5.0	

	Diameter In Centimetres At Breast Height	Ward Code	\
151	12.0	E05000141	
384	5.0	E05000139	
495	11.0	E05000131	
611	4.0	E05000130	
653	7.0	E05000136	
...
21826	7.0	E05000139	
22948	6.0	E05000138	
23266	8.0	E05000132	
23335	12.0	E05000143	
23372	8.0	E05000138	

	Ward Name	Easting	Northing	Longitude	\
151	King's Cross	530227	182706	-0.124054	
384	Kentish Town	529007	185975	-0.140440	
495	Cantelowes	529202	184717	-0.138094	
611	Camden Town with Primrose Hill	527962	184050	-0.156202	
653	Haverstock	528072	184723	-0.154369	
...
21826	Kentish Town	528962	185826	-0.141137	
22948	Holborn and Covent Garden	529976	181160	-0.128249	
23266	Fortune Green	524568	185347	-0.204661	

23335	St Pancras and Somers Town	529662	183417	-0.131941
23372	Holborn and Covent Garden	530742	182119	-0.116856

	Latitude	Location
151	51.528257	(51.528257, -0.124054)
384	51.557913	(51.557913, -0.14044)
495	51.546569	(51.546569, -0.138094)
611	51.540853	(51.540853, -0.156202)
653	51.546881	(51.546881, -0.154369)
...
21826	51.556589	(51.556589, -0.141137)
22948	51.514422	(51.514422, -0.128249)
23266	51.553268	(51.553268, -0.204661)
23335	51.534780	(51.53478, -0.131941)
23372	51.522865	(51.522865, -0.116856)

[76 rows x 17 columns]

```
[181]: # Confirmed how many rows we have
trees[mask].shape
```

```
[181]: (76, 17)
```

0.7.4 8.4 Observations

For the Identifiers in the Trees data and Environmental Dataset, there were 23 rows that didn't contain matching data.

For the Environmental data and the Trees data, I didn't identify any mismatches between the data in terms of the Identifier.

For the Trees data and Common names data, I identified 76 rows that didn't match in terms of the Scientific name.

1 END OF NOTEBOOK

```
[ ]:
```