

Analytics and Systems of Big Data Practice

PROBLEM SET 3

Bhaavana Thumu - CED17I021

Question 1

Problem statement :

1. Suppose that the data for analysis includes the attribute age. The age values for the data tuples are (in increasing order) 13, 15, 16, 16, 19, 20, 20, 21, 22, 22, 25, 25, 25, 25, 30, 33, 33, 35, 35, 35, 35, 36, 40, 45, 46, 52, 70.

- (a) Use min-max normalization to transform the values of age to the range [0:1].
- (b) Use z-score normalization to transform the values of age.
- (c) Use normalization by decimal scaling to transform the values of age such that the transformed value is less than 1.

Logic used :

Compute the mean and std dev using the statistics library and compute the min-max normalization and z-score normalization using the obtained values. Compute the power of 10 just higher than the largest element and compute normalization by decimal using that.

Packages used :

statistics - to compute the mean and std dev

Code snippets :

```
1 import statistics
2
3 age = [13, 15, 16, 16, 19, 20, 20, 21, 22, 22, 25, 25, 25, 25, 30, 33, 33, 35, 35, 35, 35, 36, 40, 45, 46, 52, 70]
4
5 min_max_normalization = []
6 z_score_normalization = []
7 decimal_scaling = []
8
9 mean = statistics.mean(age)
10 std_dev = statistics.stdev(age)
11
12 div = 1
13 while(div < age[len(age)-1]):
14     div = div*10
15
16 deno = age[(len(age)-1)] - age[0]
17
18 for i in range(len(age)):
19     min_max_normalization.append(round((age[i]-age[0])/deno, 6))
20     z_score_normalization.append(round((age[i]-mean)/std_dev, 6))
21     decimal_scaling.append(round(age[i]/div, 6))
22
23 print("min max normalization : ", min_max_normalization, "\n")
24 print("z-score normalization : ", z_score_normalization, "\n")
25 print("normalization by decimal scaling : ", decimal_scaling, "\n")
26
```

Detailed output :

```
dell@bhaavanaa:~/semester_8/big_data/lab/lab_3$ python3 Q1.py
min max normalization : [0.0, 0.035088, 0.052632, 0.052632, 0.105263, 0.122807, 0.122807, 0.140351, 0.157895,
0.157895, 0.210526, 0.210526, 0.210526, 0.210526, 0.298246, 0.350877, 0.350877, 0.385965, 0.385965, 0.385965,
0.385965, 0.403509, 0.473684, 0.561404, 0.578947, 0.684211, 1.0]

z-score normalization : [-1.310678, -1.156144, -1.078877, -1.078877, -0.847076, -0.769809, -0.769809, -0.6925
42, -0.615275, -0.615275, -0.383474, -0.383474, -0.383474, -0.383474, 0.002862, 0.234663, 0.234663, 0.389197,
0.389197, 0.389197, 0.389197, 0.466464, 0.775532, 1.161868, 1.239135, 1.702737, 3.093545]

normalization by decimal scaling : [0.13, 0.15, 0.16, 0.16, 0.19, 0.2, 0.2, 0.21, 0.22, 0.22, 0.25, 0.25, 0.2
5, 0.25, 0.3, 0.33, 0.33, 0.35, 0.35, 0.35, 0.35, 0.36, 0.4, 0.45, 0.46, 0.52, 0.7]

dell@bhaavanaa:~/semester_8/big_data/lab/lab_3$
```

Question 2

Problem statement :

Use the given dataset and perform the operations listed below.

a. Sort the attribute “Total Volume” in the given dataset and distribute the data into equal sized/frequency bins. Let the number of bins be 250. Smooth the sorted data by

(i) bin-means

(ii) bin-medians

(iii) bin-boundaries

b. The dataset represents weekly retail scan data for National retail volume (units) and price. However, the company is interested in knowing the monthly (total per month) and annual sales (total per year), rather than the total per week. So, reduce the data accordingly.

c. Summarize the number of missing values for each attribute

d. Populate data for the missing values of the attribute= “Average Price” by averaging all the values of the “Avg Price” attribute that fall under the same “REGION” attribute value.

e. Discretize the attribute= “Date” using concept hierarchy into {Old, New, Recent} (Consider 2015,2016 : Old, 2017: New, 2018: Recent).

Logic used :

- a) Take the total volume column in a separate list and calculate the bin size given the number of bins. Sort the list. Initialize the bins for mean, median and boundary. For mean bin, compute the mean of the elements of the bin and replace all the elements of the bin with the mean. Similarly for median also. For the boundaries bin, for each element in the bin, check whether it is closer to the first element of the bin or the last element and replace it with the closer one.

-
- b) Based on the month-year and the region, find the avg of the avg price and the sum of the other quantities and reduce the data accordingly.
 - c) We must count the missing values for every column of the dataset. This can be done by checking if the element is empty or if there is “not available”. Empty elements can be checked by string comparison with ‘’, while not available can be checked for a few columns by seeing if the element is a floating point number or not. If it is a floating point number, then the value is available, otherwise the count of the respective column is incremented. Display the counts of the not available columns.
 - d) Check which value is missing and replace it with the average price by calculating for the same region.
 - e) Check the year from the year column and update the date column accordingly.
2015 and 2016 - old; 2017 - new; 2018 - recent.

Packages used :

csv - to read the dataset

math - to compute ceil and floor

numpy - for working with arrays

Code snippets and output :

Reading the dataset into a variable.

```
1  ### QUESTION 2 - Use the given dataset and perform the operations listed below.
2
3
4  import csv                                # import the required libraries
5  import math
6  import numpy as np
7
8
9  with open("Avocado Dataset.csv", 'r') as f:    # open the dataset in read mode
10     data = list(csv.reader(f, delimiter = ",")) # read the dataset into the variable data
11
```

a)

```
13 ## a) Sort the attribute "Total Volume" in the given dataset and distribute the data into equal sized/frequency bins.
14 ## Let the number of bins be 250. Smooth the sorted data by (i)bin-means, (ii) bin-medians, (iii) bin-boundaries
15
16 tvol = [] # initialize list to store the total volume column
17 for i in range(1, len(data)): # convert the elements under the total volume column to float
18     tvol.append(float(data[i][2]))
19
20 total_volume = np.array(tvol) # convert the list to float
21 total_volume = np.sort(total_volume) # sort the total_volume
22
23 no_of_bins = 250 # given that the no. of bins is 250
24 bin_size = math.ceil(len(total_volume) / no_of_bins) # so, bin size can be found by total no. of elements / no. of bins
25
26 bin1 = np.zeros((no_of_bins, bin_size)) # create bin1 for mean
27 bin2 = np.zeros((no_of_bins, bin_size)) # create bin2 for median
28 bin3 = np.zeros((no_of_bins, bin_size)) # create bin3 for boundaries
29
30 for i in range(0, no_of_bins*bin_size, bin_size): # bin mean
31     k = int(i/bin_size) # required for storing the index of the bin
32     mean = sum(total_volume[i:i+bin_size])/bin_size # obtaining the mean of the elements present in the bin
33     for j in range(bin_size): # replace the elements of the bin with the mean
34         bin1[k,j] = round(mean, 6)
35 print("Bin Mean: \n",bin1) # print the mean bin
36
37 for i in range(0, no_of_bins*bin_size, bin_size): # bin median
38     k = int(i/bin_size) # required for storing the index of the bin
39     for j in range(bin_size): # replace the elements of the bin with the median
40         bin2[k,j] = total_volume[i+math.floor(bin_size/2)] # compute the median
41 print("Bin Median: \n",bin2) # print the median bin
42
43 for i in range(0, no_of_bins*bin_size, bin_size): # bin boundaries
44     k = int(i/bin_size) # required for storing the index of the bin
45     for j in range(bin_size): # replace the elements of the bin with the closest boundary element
46         if (total_volume[i+j]-total_volume[i]) < (total_volume[i+bin_size-1]-total_volume[i+j]):
47             bin3[k,j] = total_volume[i] # if the element is closer to the start element
48         else:
49             bin3[k,j] = total_volume[i+bin_size-1] # if the element is closer to the last element
50 print("Bin Boundaries: \n",bin3) # print the boundaries bin
```


b)

```
53 ## b) The dataset represents weekly retail scan data for National retail volume (units) and price.
54 ## However, the company is interested in knowing the monthly (total per month) and annual sales (total per year),
55 ## rather than the total per week. So, reduce the data accordingly.
56
57 date = data[1][0][-7:] # for obtaining the month-year from date column - first date of dataset
58 region = data[1][12] # for obtaining region - first region of dataset
59 c = 0 # for count - calculation of avg price per month
60
61 list_date = [] # initialize the required lists
62 list_avg_price = []
63 list_total_volume = []
64 list_4046 = []
65 list_4225 = []
66 list_4770 = []
67 list_total_bags = []
68 list_small_bags = []
69 list_large_bags = []
70 list_xlarge_bags = []
71 list_region = []
72
73 sum_avg_price = 0 # initialize the sum variables to zero
74 sum_total_volume = 0
75 sum_4046 = 0
76 sum_4225 = 0
77 sum_4770 = 0
78 sum_total_bags = 0
79 sum_small_bags = 0
80 sum_large_bags = 0
81 sum_xlarge_bags = 0
82
83 for i in range(1, len(data)):
84     if(date == data[i][0][-7:] and region == data[i][12]): # if the month-year and region match with initialized values
85         d = data[i][1] # for checking if the avg_price is float or Nan
86         if(d.replace('.', '', 1).isdigit() == True):
87             sum_avg_price += float(data[i][1]) # if float - add to the sum_avg_price, else ignore that cell
88             c += 1 # and increment the count
89             sum_total_volume += float(data[i][2]) # update the respective sum values
90             sum_4046 += float(data[i][3])
91             sum_4225 += float(data[i][4])
92             sum_4770 += float(data[i][5])
93             sum_total_bags += float(data[i][6])
94             sum_small_bags += float(data[i][7])
95             sum_large_bags += float(data[i][8])
96             sum_xlarge_bags += float(data[i][9])
97         else:
98             list_date.append(date) # if month-year doesnt match with the initialized values
99             if(c != 0): # append the computed data to the list initialized
100                 list_avg_price.append(sum_avg_price/c) # if all values of avg_price is Nan
101             else: # avg price is computed
102                 list_avg_price.append(0) # else avg_price for the month = 0
103             list_total_volume.append(sum_total_volume) # append the other computed sums to the respective lists
104             list_4046.append(sum_4046)
105             list_4225.append(sum_4225)
106             list_4770.append(sum_4770)
107             list_total_bags.append(sum_total_bags)
108             list_small_bags.append(sum_small_bags)
109             list_large_bags.append(sum_large_bags)
110             list_xlarge_bags.append(sum_xlarge_bags)
111             list_region.append(region)
112             date = data[i][0][-7:] # initialize the modified values, as it is different from previous one
113             region = data[i][12]
114             d = data[i][1]
115             if(d.replace('.', '', 1).isdigit() == True):
116                 sum_avg_price = float(data[i][1])
117                 c = 1
118             else:
119                 sum_avg_price = 0
120                 c = 0
121             sum_total_volume = float(data[i][2])
122             sum_4046 = float(data[i][3])
123             sum_4225 = float(data[i][4])
124             sum_4770 = float(data[i][5])
125             sum_total_bags = float(data[i][6])
126             sum_small_bags = float(data[i][7])
127             sum_large_bags = float(data[i][8])
128             sum_xlarge_bags = float(data[i][9])
```

```

130 print("Date      avg_price  total_volume  4046  4225  4770  total_bags  small_bags  large_bags  xlarge_bags  list_region")
131 for i in range(len(list_date)):
132     print(list_date[i], " ", round(list_avg_price[i], 2), " ", round(list_total_volume[i], 2), " ", round(list_4046[i], 2), " ",
133           round(list_4225[i], 2), " ", round(list_4770[i], 2), " ", round(list_total_bags[i], 2), " ", round(list_small_bags[i], 2), " ",
134           round(list_large_bags[i], 2), " ", round(list_xlarge_bags[i], 2), " ", list_region[i])
135

```

c)

```

137 ## c) Summarize the number of missing values for each attribute
138
139 col0 = 0 # initialize the count for all columns to 0
140 col1 = 0
141 col2 = 0
142 col3 = 0
143 col4 = 0
144 col5 = 0
145 col6 = 0
146 col7 = 0
147 col8 = 0
148 col9 = 0
149 col10 = 0
150 col11 = 0
151 col12 = 0
152
153 for i in range(1, len(data)):
154     d0 = data[i][0] # store the required values for checking if the string is float
155     d1 = data[i][1]
156     d2 = data[i][2]
157     d3 = data[i][3]
158     d4 = data[i][4]
159     d5 = data[i][5]
160     d6 = data[i][6]
161     d7 = data[i][7]
162     d8 = data[i][8]
163     d9 = data[i][9]
164     if(d0 == ''): # if the values are not available, increment count for that column
165         print(d0)
166         col0+=1
167     elif(d1 == '' or d1.replace('.', '', 1).isdigit() == False):
168         col1+=1
169     elif(d2 == '' or d2.replace('.', '', 1).isdigit() == False):
170         col2+=1
171     elif(d3 == '' or d3.replace('.', '', 1).isdigit() == False):
172         col3+=1
173     elif(d4 == '' or d4.replace('.', '', 1).isdigit() == False):
174         col4+=1
175     elif(d5 == '' or d5.replace('.', '', 1).isdigit() == False):
176         col5+=1
177     elif(d6 == '' or d6.replace('.', '', 1).isdigit() == False):
178         col6+=1
179     elif(d7 == '' or d7.replace('.', '', 1).isdigit() == False):
180         col7+=1
181     elif(d8 == '' or d8.replace('.', '', 1).isdigit() == False):
182         col8+=1
183     elif(d9 == '' or d9.replace('.', '', 1).isdigit() == False):
184         col9+=1
185     elif(data[i][10] == ''):
186         col10+=1
187     elif(data[i][11].isdigit() == False):
188         col11+=1
189     elif(data[i][12] == ''):
190         col12+=1
191
192 print("The count of the missing values are-") # print the final counts for the columns
193 print("Date - ", col0)
194 print("Average price - ", col1)
195 print("Total volume - ", col2)
196 print("4046 - ", col3)
197 print("4225 - ", col4)
198 print("4770 - ", col5)
199 print("Total bags - ", col6)
200 print("Small bags - ", col7)
201 print("Large bags - ", col8)
202 print("XLarge bags - ", col9)
203 print("Type - ", col10)
204 print("Year - ", col11)
205 print("Region - ", col12)
206

```

d)

```
207
208 ## d) Populate data for the missing values of the attribute= "Average Price" by averaging
209 ## all the values of the "Avg Price" attribute that fall under the same "REGION" attribute value.
210 can = 0
211 for i in range(1, len(data)):
212     d = data[i][1]
213     if(data[i][1] == '' or d.replace('.', '', 1).isdigit() == False): # if it is missing value
214         s = 0
215         count = 0
216         for j in range(1, len(data)):
217             d = data[j][1]
218             if(data[j][12] == data[i][12] and d.replace('.', '', 1).isdigit() == True): # if it is a float value
219                 s = s + float(data[j][1])
220                 count+=1
221             data[i][1] = str(round(s/count, 6))
222 # print(data)
223
```

e)

```
224
225 ## e) Discretize the attribute= "Date" using concept hierarchy into {Old, New, Recent}
226 ## (Consider 2015,2016 : Old, 2017: New, 2018: Recent).
227
228 for i in range(1, len(data)):
229     if((data[i][11] == '2015') or (data[i][11] == '2016')): # 2015 or 2016 - Old
230         data[i][0] = "Old"
231     elif(data[i][11] == '2017'):
232         data[i][0] = "New"
233     elif(data[i][11] == '2018'):
234         data[i][0] = "Recent"
235 # print(data)
236
```

Detailed output :

a)

```
dell@bhaavana:~/semester_8/big_data/lab/lab_3$ python3 Q2.py
Bin Mean:
[[7.35713014e+02 7.35713014e+02 7.35713014e+02 ... 7.35713014e+02
 7.35713014e+02 7.35713014e+02]
 [1.03670205e+03 1.03670205e+03 1.03670205e+03 ... 1.03670205e+03
 1.03670205e+03 1.03670205e+03]
 [1.17840082e+03 1.17840082e+03 1.17840082e+03 ... 1.17840082e+03
 1.17840082e+03 1.17840082e+03]
 ...
 [1.35563292e+07 1.35563292e+07 1.35563292e+07 ... 1.35563292e+07
 1.35563292e+07 1.35563292e+07]
 [3.11970276e+07 3.11970276e+07 3.11970276e+07 ... 3.11970276e+07
 3.11970276e+07 3.11970276e+07]
 [3.89735224e+07 3.89735224e+07 3.89735224e+07 ... 3.89735224e+07
 3.89735224e+07 3.89735224e+07]]
Bin Median:
[[7.74200000e+02 7.74200000e+02 7.74200000e+02 ... 7.74200000e+02
 7.74200000e+02 7.74200000e+02]
 [1.03500000e+03 1.03500000e+03 1.03500000e+03 ... 1.03500000e+03
 1.03500000e+03 1.03500000e+03]
 [1.17595000e+03 1.17595000e+03 1.17595000e+03 ... 1.17595000e+03
 1.17595000e+03 1.17595000e+03]
 ...
 [8.38991804e+06 8.38991804e+06 8.38991804e+06 ... 8.38991804e+06
 8.38991804e+06 8.38991804e+06]
 [3.13460915e+07 3.13460915e+07 3.13460915e+07 ... 3.13460915e+07
 3.13460915e+07 3.13460915e+07]
 [3.73523606e+07 3.73523606e+07 3.73523606e+07 ... 3.73523606e+07
 3.73523606e+07 3.73523606e+07]]
Bin Boundaries:
[[8.45600000e+01 8.45600000e+01 8.45600000e+01 ... 9.34950000e+02
 9.34950000e+02 9.34950000e+02]
 [9.36690000e+02 9.36690000e+02 9.36690000e+02 ... 1.11744000e+03
 1.11744000e+03 1.11744000e+03]
 [1.11847000e+03 1.11847000e+03 1.11847000e+03 ... 1.23327000e+03
 1.23327000e+03 1.23327000e+03]
 ...
 [7.36092584e+06 7.36092584e+06 7.36092584e+06 ... 2.80125209e+07
 2.80125209e+07 2.80125209e+07]
 [2.80413354e+07 2.80413354e+07 2.80413354e+07 ... 3.39939313e+07
 3.39939313e+07 3.39939313e+07]
 [3.41267310e+07 3.41267310e+07 3.41267310e+07 ... 6.25056465e+07
 6.25056465e+07 6.25056465e+07]]
```

b)

Date	avg_price	total_volume	4046	4225	4770	total_bags	small_bags	large_bags	xlarge_bags	list_region
12-2015	1.17	316325.97	3637.72		280219.74	389.57	32158.94	31731.3	427.64	0.0 Albany
11-2015	1.29	399712.89	5220.57		354710.63	377.99	39404.3	38110.92	1293.38	0.0 Albany
10-2015	1.31	284678.47	5617.61		242911.88	436.75	35712.23	34185.58	1526.65	0.0 Albany
09-2015	1.18	351846.63	4898.09		314481.81	688.13	32578.6	30806.03	1772.57	0.0 Albany
08-2015	1.26	452063.74	3056.39		393958.99	2285.14	52703.22	52147.26	555.96	0.0 Albany
07-2015	1.19	436681.53	3232.55		341292.52	13345.43	78811.03	78118.29	659.41	33.33 Albany
06-2015	1.26	406758.56	2975.53		288856.1	14611.4	106315.53	98631.89	1683.64	0.0 Albany
05-2015	1.26	486077.58	5893.49		371292.23	606.47	109085.39	107007.23	2078.16	0.0 Albany
04-2015	1.18	194376.61	3475.84		135735.07	252.03	54913.67	51987.45	2822.05	104.17 Albany
03-2015	1.06	253294.82	7097.51		199387.93	671.51	46137.87	44264.29	1873.58	0.0 Albany
02-2015	1.03	209370.24	4786.31		164230.65	715.99	39637.29	38263.95	1373.34	0.0 Albany
01-2015	1.17	171727.14	5677.87		124664.24	476.93	40908.1	38977.41	1930.69	0.0 Albany
12-2015	1.03	1492887.31	1151827.8		102813.3	858.68	237387.53	171472.45	65898.84	16.24 Atlanta
11-2015	1.05	1782686.38	1324982.34		184527.78	2005.99	271250.27	178804.14	92388.9	57.23 Atlanta
10-2015	1.02	1560107.28	987254.98		308167.41	1839.72	262845.17	124746.83	138098.34	0.0 Atlanta
09-2015	1.01	1776122.48	1189387.45		280795.17	3656.39	302283.47	150356.17	151796.33	130.97 Atlanta
08-2015	1.1	2136930.98	1604475.77		222002.1	11176.36	299276.75	225687.35	73433.91	155.49 Atlanta
07-2015	1.05	1964257.95	1527930.43		184554.11	4406.04	247367.37	186922.42	60420.2	24.75 Atlanta
06-2015	1.05	2006788.3	1622601.32		108045.6	2705.36	273436.02	207151.49	66027.11	257.42 Atlanta
05-2015	1.05	2643449.02	2259285.26		112875.34	4771.35	266517.07	174711.62	91753.82	51.63 Atlanta
04-2015	1.07	1713804.48	1452438.41		78081.52	2508.67	180775.88	115422.56	65218.6	134.72 Atlanta
03-2015	1.06	2178237.74	1837895.22		93552.21	1461.35	245328.96	170748.97	74526.28	53.71 Atlanta
02-2015	1.04	1929355.12	1664218.0		84491.05	1695.44	178950.63	88749.54	90201.09	0.0 Atlanta
01-2015	1.08	1713388.05	1460358.72		80199.55	2323.27	170506.51	67975.86	102530.65	0.0 Atlanta
12-2015	1.12	2508152.38	180721.41		1682132.93	86517.62	558780.42	550478.96	8301.46	0.0 BaltimoreWashington
11-2015	1.1	3409338.23	239445.04		2319095.8	191132.11	659665.28	639415.26	20250.02	0.0 BaltimoreWashington
10-2015	0	2973499.99	182937.16		1942915.16	199004.9	648642.77	627572.19	21070.58	0.0 BaltimoreWashington
09-2015	1.18	3154167.4	193531.9		2099436.81	214805.55	646393.14	627065.19	19322.41	5.54 BaltimoreWashington
08-2015	1.17	4023538.05	304509.68		2647023.62	220756.58	851248.17	835405.06	15375.48	467.63 BaltimoreWashington
07-2015	1.16	3320350.81	231034.01		2107893.23	202054.17	779369.4	760946.71	18038.68	384.01 BaltimoreWashington
06-2015	1.16	3474459.48	250041.31		2118852.39	195128.4	910437.38	894430.55	14805.59	1201.24 BaltimoreWashington
05-2015	1.21	4400870.62	347100.55		2677534.69	262410.26	1113825.12	1080239.65	33145.83	439.64 BaltimoreWashington
04-2015	1.25	3036398.78	249876.52		1783605.74	166205.69	836710.83	812136.01	23598.43	976.39 BaltimoreWashington
03-2015	1.2	3594133.8	290981.9		2185402.72	215246.85	902502.33	880393.62	22108.71	0.0 BaltimoreWashington
02-2015	1.13	3239515.1	265122.87		2028515.72	229601.46	716275.05	697530.1	18744.95	0.0 BaltimoreWashington
01-2015	1.17	2808932.8	205091.25		1753502.19	162587.02	687752.34	667629.63	20122.71	0.0 BaltimoreWashington
12-2015	0.93	274170.13	118850.06		21030.26	23832.94	110456.87	109997.94	21.56	437.37 Boise
11-2015	1.08	280026.23	136368.82		27142.76	31910.66	84603.99	83701.2	132.76	770.03 Boise
10-2015	1.08	254466.11	120404.43		63396.93	28293.95	42370.8	41584.6	690.11	96.09 Boise
09-2015	0.99	266814.15	171707.73		37142.24	24888.58	33075.6	32614.74	433.41	27.45 Boise
08-2015	1.1	323926.09	227066.08		36104.35	28179.46	32576.2	32507.58	62.22	6.4 Boise
07-2015	1.09	291506.28	210678.46		23981.35	26954.12	29892.35	27604.59	2287.76	0.0 Boise
06-2015	1.05	323936.18	334330.38		43435.08	38552.38	36310.04	36944.44	6.76	366.34 Boise

c)

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The count of the missing values are-
Date - 0
Average price - 48
Total volume - 0
4046 - 0
4225 - 0
4770 - 0
Total bags - 0
Small bags - 0
Large bags - 0
XLarge bags - 0
Type - 0
Year - 0
Region - 0

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d)

'27-12-2015'	'1.33'	'64236.62'	'1036.74'	'54454.85'	'48.16'	'8696.87'	'8603.62'	'93.25'	'0'	'conventional'	'2015'	'Albany'
'20-12-2015'	'1.35'	'54876.98'	'674.26'	'44638.81'	'58.33'	'9505.56'	'9488.07'	'97.49'	'0'	'conventional'	'2015'	'Albany'
'13-12-2015'	'0.93'	'118220.22'	'794.7'	'109149.67'	'130.5'	'8145.35'	'8042.21'	'103.14'	'0'	'conventional'	'2015'	'Albany'
'06-12-2015'	'1.08'	'78992.15'	'1132'	'71976.41'	'72.58'	'5811.16'	'5677.4'	'133.76'	'0'	'conventional'	'2015'	'Albany'
'29-11-2015'	'1.29'	'51039.6'	'941.48'	'43838.39'	'75.78'	'6183.95'	'5986.26'	'197.69'	'0'	'conventional'	'2015'	'Albany'
'22-11-2015'	'1.570755'	'55979.78'	'1184.27'	'48067.99'	'43.61'	'6683.91'	'6556.47'	'127.44'	'0'	'conventional'	'2015'	'Albany'
'15-11-2015'	'1.570755'	'83453.76'	'1368.92'	'73672.72'	'93.26'	'8318.86'	'8196.81'	'122.05'	'0'	'conventional'	'2015'	'Albany'
'08-11-2015'	'1.570755'	'109428.33'	'703.75'	'101815.36'	'80'	'6829.22'	'6266.85'	'562.37'	'0'	'conventional'	'2015'	'Albany'
'01-11-2015'	'1.570755'	'99811.42'	'1022.15'	'87315.57'	'85.34'	'11388.36'	'11104.53'	'283.83'	'0'	'conventional'	'2015'	'Albany'
'25-10-2015'	'1.570755'	'74338.76'	'842.4'	'64757.44'	'113'	'8625.92'	'8061.47'	'564.45'	'0'	'conventional'	'2015'	'Albany'
'18-10-2015'	'1.570755'	'84843.44'	'924.86'	'75595.85'	'117.07'	'8205.66'	'7877.86'	'327.8'	'0'	'conventional'	'2015'	'Albany'
'11-10-2015'	'1.570755'	'64489.17'	'1582.03'	'52677.92'	'105.32'	'10123.9'	'9866.27'	'257.63'	'0'	'conventional'	'2015'	'Albany'
'04-10-2015'	'1.31'	'61007.1'	'2268.32'	'49880.67'	'101.36'	'8756.75'	'8379.98'	'376.77'	'0'	'conventional'	'2015'	'Albany'
'27-09-2015'	'0.99'	'106803.39'	'1204.88'	'99409.21'	'154.84'	'6034.46'	'5888.87'	'145.59'	'0'	'conventional'	'2015'	'Albany'
'20-09-2015'	'1.33'	'69759.01'	'1028.03'	'59313.12'	'150.5'	'9267.36'	'8489.1'	'778.26'	'0'	'conventional'	'2015'	'Albany'
'13-09-2015'	'1.28'	'76111.27'	'985.73'	'65696.86'	'142'	'9286.68'	'8665.19'	'621.49'	'0'	'conventional'	'2015'	'Albany'
'06-09-2015'	'1.11'	'99172.96'	'879.45'	'90062.62'	'240.79'	'7990.1'	'7762.87'	'272.23'	'0'	'conventional'	'2015'	'Albany'
'30-08-2015'	'1.07'	'105693.84'	'689.01'	'94362.67'	'335.43'	'10306.73'	'10218.93'	'87.8'	'0'	'conventional'	'2015'	'Albany'
'23-08-2015'	'1.34'	'79992.09'	'733.16'	'67933.79'	'444.78'	'10880.36'	'10745.79'	'134.57'	'0'	'conventional'	'2015'	'Albany'
'16-08-2015'	'1.33'	'80043.78'	'539.65'	'68666.01'	'394.9'	'10443.22'	'10297.68'	'145.54'	'0'	'conventional'	'2015'	'Albany'
'09-08-2015'	'1.12'	'111140.93'	'584.63'	'100961.46'	'368.95'	'9225.89'	'9116.34'	'109.55'	'0'	'conventional'	'2015'	'Albany'
'02-08-2015'	'1.45'	'75133.1'	'509.94'	'62035.06'	'741.08'	'11847.02'	'11768.52'	'78.5'	'0'	'conventional'	'2015'	'Albany'
'26-07-2015'	'1.11'	'106757.1'	'648.75'	'91949.05'	'966.61'	'13192.69'	'13061.53'	'131.16'	'0'	'conventional'	'2015'	'Albany'
'19-07-2015'	'1.26'	'96617.'	'1042.1'	'82049.4'	'2238.02'	'11287.48'	'11103.49'	'183.99'	'0'	'conventional'	'2015'	'Albany'
'12-07-2015'	'1.05'	'124055.31'	'672.25'	'94693.52'	'4257.64'	'24431.9'	'24290.08'	'108.49'	'33.33'	'conventional'	'2015'	'Albany'
'05-07-2015'	'1.35'	'109252.12'	'869.45'	'72600.55'	'5883.16'	'29898.96'	'29663.19'	'235.77'	'0'	'conventional'	'2015'	'Albany'
'28-06-2015'	'1.37'	'89534.81'	'664.23'	'57545.79'	'4662.71'	'26662.08'	'26311.76'	'350.32'	'0'	'conventional'	'2015'	'Albany'
'21-06-2015'	'1.27'	'104849.39'	'804.01'	'76688.55'	'5481.18'	'21875.65'	'21662'	'213.65'	'0'	'conventional'	'2015'	'Albany'
'14-06-2015'	'1.32'	'89631.3'	'850.58'	'55400.94'	'4377.19'	'29002.59'	'28343.14'	'659.45'	'0'	'conventional'	'2015'	'Albany'

e)

'Old'	'1.25'	'50809.3'	'6329.81'	'28751.94'	'572.18'	'15155.37'	'5948.7'	'9206.67'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.1'	'41244.99'	'4759.58'	'9207.54'	'14337.18'	'12940.69'	'7111.59'	'5829.1'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.09'	'52358.11'	'5422.01'	'17237.12'	'21948.79'	'7750.19'	'5167.45'	'2582.74'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.36'	'38767.82'	'2907.01'	'25699'	'262.86'	'9898.95'	'5774.63'	'4124.32'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.36'	'37885.9'	'2655.44'	'25121.02'	'177.35'	'19932.09'	'5036.29'	'4895.8'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.31'	'21561.29'	'3204.95'	'7746.53'	'111.55'	'10490.26'	'8558.96'	'1939.3'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.29'	'20698.71'	'2278.61'	'6800.76'	'58.05'	'11561.28'	'10538.62'	'1022.66'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.41'	'17528.02'	'2983.3'	'6722.24'	'53.08'	'7769.4'	'7268.06'	'501.34'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.47'	'15705.35'	'2731.96'	'6730.01'	'52.95'	'6190.43'	'6185.08'	'5.35'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.3'	'18611.51'	'1716.44'	'8405.31'	'60.03'	'8429.73'	'8365.7'	'64.03'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.32'	'18916.59'	'2653.06'	'5464.63'	'14.37'	'10784.53'	'10704.67'	'79.86'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.36'	'16034.56'	'2064.7'	'5044.66'	'7.17'	'8918.03'	'8879.53'	'38.5'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.48'	'14413.74'	'1308.04'	'4865.24'	'4.77'	'8235.69'	'7090.72'	'1144.97'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.39'	'15805.58'	'1483.39'	'4975.83'	'38.02'	'9308.34'	'6330.13'	'2978.21'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.49'	'13712.44'	'1451'	'4199.43'	'37.92'	'8024.09'	'7913.5'	'110.59'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.64'	'11626.56'	'1618.9'	'4032.03'	'2.34'	'5973.29'	'5957.67'	'15.62'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.47'	'12171.04'	'1554.09'	'4656.41'	'9.36'	'5951.18'	'5852.36'	'98.82'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.45'	'13237.48'	'1912.6'	'4779.17'	'25.72'	'6519.99'	'6447.25'	'72.74'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.37'	'12647.9'	'1591.6'	'4070.29'	'23.41'	'6962.6'	'6780.56'	'182.04'	'0'	'organic'	'2016'	'WestTexNewMexico'
'Old'	'1.58'	'9667.05'	'1020.39'	'2703.11'	'16.41'	'5047.14'	'4651.16'	'395.98'	'0'	'organic'	'2016'	'WestTexNewMexico'
'New'	'1.46'	'3463.85'	'18.12'	'190.16'	'0'	'3247.57'	'3247.57'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.58'	'3694.13'	'31.68'	'327.39'	'0'	'3335.06'	'3335.06'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.43'	'3513.77'	'37.46'	'209.3'	'0'	'3267.01'	'3267.01'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.45'	'3779.98'	'18.04'	'262.14'	'0'	'3499.8'	'3499.8'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.44'	'3577.04'	'118.55'	'306.55'	'0'	'3151.94'	'3151.94'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.57'	'2841.29'	'27.75'	'182.15'	'0'	'2631.39'	'2631.39'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.75'	'2506.38'	'0'	'252.98'	'0'	'2253.4'	'2250.07'	'3.33'	'0'	'organic'	'2017'	'Albany'
'New'	'1.71'	'2664.62'	'0'	'245.71'	'0'	'2418.91'	'2418.91'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.5'	'3425.86'	'3.56'	'64.14'	'0'	'3358.16'	'3358.16'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.38'	'5583.79'	'12.99'	'180.62'	'0'	'5390.18'	'5390.18'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.49'	'4337.67'	'3.52'	'268.83'	'0'	'4065.32'	'4061.99'	'3.33'	'0'	'organic'	'2017'	'Albany'
'New'	'1.47'	'4522.84'	'18.71'	'237.43'	'0'	'4266.7'	'4266.7'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.4'	'5229.43'	'23.15'	'341.52'	'0'	'4864.76'	'4864.76'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.59'	'3423.95'	'31.21'	'150.24'	'0'	'3242.51'	'3242.51'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.42'	'3627.18'	'56.82'	'95.31'	'0'	'3475.05'	'3475.05'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.55'	'3431.2'	'30.32'	'215.1'	'0'	'3185.78'	'3185.78'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.78'	'3139.84'	'45.5'	'196.27'	'0'	'2898.07'	'2894.74'	'3.33'	'0'	'organic'	'2017'	'Albany'
'New'	'2'	'2022'	'112.42'	'254.49'	'0'	'1655.09'	'1655.09'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.91'	'2525.28'	'44.15'	'210.54'	'0'	'2270.59'	'2270.59'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.86'	'2584.08'	'61.21'	'143.82'	'0'	'2379.05'	'2379.05'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.9'	'2259.92'	'42.86'	'207.29'	'0'	'2009.77'	'2009.77'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.98'	'2576.49'	'39.31'	'331.89'	'0'	'2205.29'	'2205.29'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.67'	'2503.82'	'55.26'	'92.99'	'0'	'2355.57'	'2355.57'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.42'	'4233.61'	'93.39'	'93.39'	'0'	'4046.83'	'4046.83'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.87'	'2889.03'	'155.1'	'274.95'	'0'	'2458.98'	'2458.98'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'2'	'1883.11'	'38.06'	'187.9'	'0'	'1657.14'	'1657.14'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'2.03'	'2268.86'	'59.41'	'278.04'	'0'	'1931.41'	'1931.41'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'2.13'	'2089.01'	'142.46'	'295.6'	'0'	'2460.75'	'2460.75'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'2.03'	'3185.1'	'366.52'	'266.88'	'0'	'2551.71'	'2551.71'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'2.04'	'2719.24'	'21.33'	'248.87'	'0'	'2449.04'	'2449.04'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.63'	'3771.39'	'54.47'	'352.85'	'0'	'3364.07'	'3364.07'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.87'	'3365.45'	'11.93'	'344.88'	'0'	'3008.64'	'3008.64'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.84'	'3184.37'	'20.33'	'385.05'	'0'	'2778.99'	'2778.99'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.57'	'4270.28'	'44.12'	'450.74'	'0'	'3775.42'	'3775.42'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.79'	'2897.23'	'89.53'	'315.14'	'0'	'2492.56'	'2492.56'	'0'	'0'	'organic'	'2017'	'Albany'
'New'	'1.74'	'3046.63'	'188.81'	'288.28'	'0'	'2377.54'	'2377.54'	'0'	'0'	'organic'	'2017'	'Albany'

['New', '1.03', '848944.78', '488728.09', '118607.98', '9864.53', '231744.18', '125662.18', '100565.34', '5516.66', 'conventional', '2017', 'WestTexNewMexico']
['New', '0.95', '887639.13', '473343.11', '152255.69', '24289.97', '237750.36', '111162.35', '126588.01', '0', 'conventional', '2017', 'WestTexNewMexico']
['New', '0.84', '1003016.1', '549057.13', '109380.77', '8436.06', '45413.21', '192156.03', '223256.38', '0', 'conventional', '2017', 'WestTexNewMexico']
['New', '0.86', '1023584.42', '541032.2', '87497.9', '10276.33', '384777.99', '108553.29', '216224.7', '0', 'conventional', '2017', 'WestTexNewMexico']
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