main

September 11, 2023

1 Association rule mining

In this notebook, you'll implement the basic pairwise association rule mining algorithm.

To keep the implementation simple, you will apply your implementation to a simplified dataset, namely, letters ("items") in words ("receipts" or "baskets"). Having finished that code, you will then apply that code to some grocery store market basket data. If you write the code well, it will not be difficult to reuse building blocks from the letter case in the basket data case.

1.1 Problem definition

Let's say you have a fragment of text in some language. You wish to know whether there are association rules among the letters that appear in a word. In this problem:

- Words are "receipts"
- Letters within a word are "items"

You want to know whether there are *association rules* of the form, $a \implies b$, where a and b are letters. You will write code to do that by calculating for each rule its *confidence*, $conf(a \implies b)$. "Confidence" will be another name for an estimate of the conditional probability of b given a, or $Pr[b \mid a]$.

1.2 Sample text input

Let's carry out this analysis on a "dummy" text fragment, which graphic designers refer to as the *lorem ipsum*:

veritatis et quasi architecto beatae vitae dicta sunt, explicabo. Nemo enim ipsam voluptatem, quia voluptas sit, aspernatur aut odit aut fugit, sed quia consequuntur magni dolores eos, qui ratione voluptatem sequi nesciunt, neque porro quisquam est, qui dolorem ipsum, quia dolor sit amet consectetur adipisci[ng] velit, sed quia non numquam [do] eius modi tempora inci[di]dunt, ut labore et dolore magnam aliquam quaerat voluptatem. Ut enim ad minima veniam, quis nostrum exercitationem ullam corporis suscipit laboriosam, nisi ut aliquid ex ea commodi consequatur? Quis autem vel eum iure reprehenderit, qui in ea voluptate velit esse, quam nihil molestiae consequatur, vel illum, qui dolorem eum fugiat, quo voluptas nulla pariatur?

At vero eos et accusamus et iusto odio dignissimos ducimus, qui blanditiis praesentium voluptatum deleniti atque corrupti, quos dolores et quas molestias excepturi sint, obcaecati cupiditate non provident, similique sunt in culpa, qui officia deserunt mollitia animi, id est laborum et dolorum fuga. Et harum quidem rerum facilis est et expedita distinctio. Nam libero tempore, cum soluta nobis est eligendi optio, cumque nihil impedit, quo minus id, quod maxime placeat, facere possimus, omnis voluptas assumenda est, omnis dolor repellendus. Temporibus autem quibusdam et aut officiis debitis aut rerum necessitatibus saepe eveniet, ut et voluptates repudiandae sint et molestiae non recusandae. Itaque earum rerum hic tenetur a sapiente delectus, ut aut reiciendis voluptatibus maiores alias consequatur aut perferendis doloribus asperiores repellat. 0.00

```
print("First 100 characters:\n {} ...".format(latin_text[:100]))
```

First 100 characters:

Sed ut perspiciatis, unde omnis iste natus error sit voluptatem accusantium doloremque laudantium, ...

Exercise 0 (ungraded). Look up and read the translation of *lorem ipsum!*

Data cleaning. Like most data in the real world, this dataset is noisy. It has both uppercase and lowercase letters, words have repeated letters, and there are all sorts of non-alphabetic characters. For our analysis, we should keep all the letters and spaces (so we can identify distinct words), but we should ignore case and ignore repetition within a word.

For example, the eighth word of this text is "error." As an *itemset*, it consists of the three unique letters, $\{e, o, r\}$. That is, treat the word as a set, meaning you only keep the unique letters.

This itemset has three possible *itempairs*: $\{e, o\}$, $\{e, r\}$, and $\{o, r\}$.

Since sets are unordered, note that we would regard $\{e,o\} = \{o,e\}$, which is why we say there are only three itempairs, rather than six.

Start by writing some code to help "clean up" the input.

Exercise 1 (2 points). Complete the following function, normalize_string(s). The input s is a string (str object). The function should return a new string with (a) all characters converted to lowercase and (b) all non-alphabetic, non-whitespace characters removed.

Clarification. Scanning the sample text, latin_text, you may see things that look like special cases. For instance, inci[di]dunt and [do]. For these, simply remove the non-alphabetic characters and only separate the words if there is explicit whitespace.

For instance, inci[di]dunt would become incididunt (as a single word) and [do] would become do as a standalone word because the original string has whitespace on either side. A period or comma without whitespace would, similarly, just be treated as a non-alphabetic character inside a word *unless* there is explicit whitespace. So e pluribus.unum basium would become e pluribusunum basium even though your common-sense understanding might separate pluribus and unum.

Hint. Regard as a whitespace character anything "whitespace-like." That is, consider not just regular spaces, but also tabs, newlines, and perhaps others. To detect whitespaces easily, look for a "high-level" function that can help you do so rather than checking for literal space characters.

```
In [3]: ### Define demo inputs
    demo_s_ex1 = latin_text[:100]
    print(demo_s_ex1)
```

Sed ut perspiciatis, unde omnis iste natus error sit voluptatem accusantium doloremque laudantium,

The demo included in the solution cell below should display the following output:

sed ut perspiciatis unde omnis iste natus error sit voluptatem accusantium doloremque laudantium

```
In [4]: def normalize_string(s):
    assert type (s) is str
    ###
    ### ### YOUR CODE HERE
    new_str = [c for c in s.lower() if c.isalpha() or c.isspace()]
# print(new_str)
    new_str2 = "".join(new_str)
# print(new_str2)
```

The cell below will test your solution for Exercise 1. The testing variables will be available for debugging under the following names in a dictionary format. - input_vars - Input variables for your solution. - original_input_vars - Copy of input variables from prior to running your solution. These *should* be the same as input_vars - otherwise the inputs were modified by your solution. - returned_output_vars - Outputs returned by your solution. - true_output_vars - The expected output. This *should* "match" returned_output_vars based on the question requirements - otherwise, your solution is not returning the correct output.

```
In [5]: ### test_cell_ex1
        from tester_fw.testers import Tester
        conf = {
            'case_file':'tc_1',
            'func': normalize_string, # replace this with the function defined above
            'inputs':{ # input config dict. keys are parameter names
                's':{
                     'dtype': 'str', # data type of param.
                    'check_modified':False,
                }
            },
            'outputs':{
                'output_0':{
                    'index':0,
                    'dtype':'str',
                    'check_dtype': True,
                    'check_col_dtypes': True, # Ignored if dtype is not df
                    'check_col_order': True, # Ignored if dtype is not df
                    'check_row_order': True, # Ignored if dtype is not df
                    'check_column_type': True, # Ignored if dtype is not df
                    'float tolerance': 10 ** (-6)
                }
            }
        }
        tester = Tester(conf, key=b's63L2lglDfBJpcKzxpwcyy61HyKnJNBOJX19BMyWhyo=', path='resour
        for _ in range(70):
            try:
                tester.run_test()
```

Exercise 2 (1 point). Implement the following function, get_normalized_words(s). It takes as input a string s (i.e., a str object). It should normalize s and then return a list of its words. (That is, the function should not assume that the input s is normalized yet.)

```
demo_s_ex2 = latin_text[:33]
        demo_s_ex2
Out[6]: '\nSed ut perspiciatis, unde omnis '
   The demo included in the solution cell below should display the following output:
['sed', 'ut', 'perspiciatis', 'unde', 'omnis']
In [7]: def get_normalized_words (s):
            assert type(s) is str
            ###
            ### YOUR CODE HERE
            normalized_words = ""
            normalized_words = normalize_string(s)
            print(normalized_words)
            normalized_list = []
            normalized_list = normalized_words.split()
             print(normalized_list)
            return normalized list
            ###
        # Demo:
        print(get_normalized_words(demo_s_ex2))
```

['sed', 'ut', 'perspiciatis', 'unde', 'omnis']

In [6]: ### Define demo inputs

The cell below will test your solution for Exercise 2. The testing variables will be available for debugging under the following names in a dictionary format. - input_vars - Input variables for your solution. - original_input_vars - Copy of input variables from prior to running your solution. These *should* be the same as input_vars - otherwise the inputs were modified by your solution. - returned_output_vars - Outputs returned by your solution. - true_output_vars - The expected output. This *should* "match" returned_output_vars based on the question requirements - otherwise, your solution is not returning the correct output.

```
In [8]: ### test_cell_ex2
        from tester_fw.testers import Tester
        conf = {
            'case_file':'tc_2',
            'func': get_normalized_words, # replace this with the function defined above
            'inputs':{ # input config dict. keys are parameter names
                     'dtype':'str', # data type of param.
                    'check_modified':False,
                }
            },
            'outputs':{
                'output_0':{
                    'index':0,
                    'dtype':'',
                    'check_dtype': True,
                    'check_col_dtypes': True, # Ignored if dtype is not df
                    'check_col_order': True, # Ignored if dtype is not df
                    'check_row_order': True, # Ignored if dtype is not df
                    'check_column_type': True, # Ignored if dtype is not df
                    'float_tolerance': 10 ** (-6)
                }
            }
        tester = Tester(conf, key=b's63L2lglDfBJpcKzxpwcyy61HyKnJNBOJX19BMyWhyo=', path='resour
        for _{in} range(70):
            try:
                tester.run_test()
                (input_vars, original_input_vars, returned_output_vars, true_output_vars) = te
                (input_vars, original_input_vars, returned_output_vars, true_output_vars) = te
                raise
        print('Passed! Please submit.')
```

Exercise 3 (2 points). Implement a function, make_itemsets_unstructured_text(text). The input, text, is a string of unstructured text (like the latin_text example above). Your function should get the normalized words from the text, convert the characters of each word into an itemset and then return the list of all itemsets. These output itemsets should appear in the same order as their corresponding words in the input. You may find it helpful to call get_normalized_words in your solution.

```
In [9]: ### Define demo inputs
    demo_text_ex3 = 'sed \tut, perspiciatis\n und.e omnis'
```

Passed! Please submit.

The demo included in the solution cell below should display the following output:

```
[{'d', 'e', 's'},

{'t', 'u'},

{'a', 'c', 'e', 'i', 'p', 'r', 's', 't'},

{'d', 'e', 'n', 'u'},

{'i', 'm', 'n', 'o', 's'}]
```

Because sets are unordered, different versions of Python may produce sets with whose element-ordering differs from what you see above. However, the sets themselves should be in this order in the output list, since that is the order in which the corresponding words were given.

The cell below will test your solution for Exercise 3. The testing variables will be available for debugging under the following names in a dictionary format. - input_vars - Input variables for your solution. - original_input_vars - Copy of input variables from prior to running your solution. These *should* be the same as input_vars - otherwise the inputs were modified by your solution. - returned_output_vars - Outputs returned by your solution. - true_output_vars - The expected output. This *should* "match" returned_output_vars based on the question requirements - otherwise, your solution is not returning the correct output.