4_medication_analysis

July 31, 2023

```
[1]: import json
     import os
     from sklearn.feature_extraction.text import CountVectorizer
     import pandas as pd
     import numpy as np
     from drug named entity recognition import find drugs
     import json5
     import sys
     parent_dir = os.path.abspath("..")
     if parent_dir not in sys.path:
         sys.path.append(parent_dir)
     from path import DATA_PROCESSED_DOCUMENTS_DIR
[2]: chief_complaint = "abdominal-pain"
     folder_location = os.path.join(
         DATA_PROCESSED_DOCUMENTS_DIR / chief_complaint / "black-or-african-american"
     b docs = []
     w_{docs} = []
     for filename in os.listdir(folder_location):
         file_location = os.path.join(folder_location, filename)
         if os.path.isfile(file_location):
             with open(file_location) as d:
                 try:
                     file_contents = d.read()
                     content = json.loads(file_contents)
                     b_docs.append(content)
                 except Exception as e:
                     try:
                         # pull of first and last line, gpt sometimes response with
      →a leading ```json and ends with ```
                         tmp = file_contents.splitlines(True)
```

while "{" not in tmp[0]:
 tmp = tmp[1:]
while "}" not in tmp[-1]:
 tmp = tmp[:-1]

```
tmp = "".join(tmp)
                         content = json5.loads(tmp)
                         b_docs.append(content)
                     except Exception as e:
                         # print(f"{file_location} Error: {e}")
                         pass
    folder_location = os.path.join(DATA_PROCESSED_DOCUMENTS_DIR / chief_complaint /_
      for filename in os.listdir(folder_location):
        file_location = os.path.join(folder_location, filename)
         if os.path.isfile(file_location):
             with open(file_location) as d:
                 try:
                     file_contents = d.read()
                     content = json.loads(file_contents)
                     w_docs.append(content)
                 except Exception as e:
                     try:
                         # pull of first and last line, gpt sometimes response with
      ⇔a leading ```json and ends with
                        tmp = file_contents.splitlines(True)
                         while "{" not in tmp[0]:
                             tmp = tmp[1:]
                         while "\}" not in tmp[-1]:
                             tmp = tmp[:-1]
                         tmp = "".join(tmp)
                         content = json5.loads(tmp)
                         w_docs.append(content)
                     except Exception as e:
                         # print(f"{file_location} Error: {e}")
                         pass
[3]: print(len(b_docs))
    print(len(w_docs))
    4945
    4951
[4]: b_normalized_medications = []
    for doc in b docs:
         if doc.get("medications") is not None:
            res = doc.get("medications").split(" ")
            try:
                res.remove("other")
             except ValueError:
```

```
res = find_drugs(res, is_ignore_case=True)
             b_normalized_medications.append(res)
     len(b_normalized_medications)
[4]: 4935
[5]: b_normalized_medications[:2]
[5]: [[],
      [({'name': 'Lisinopril',
         'synonyms': {'Lisinopril',
          'Lisinoprilum',
          'Lysinopril',
          'Prinivil',
          'Zestril'},
         'medline_plus_id': 'a692051',
         'nhs_url': 'https://www.nhs.uk/medicines/lisinopril',
         'wikipedia_url': 'https://en.wikipedia.org/wiki/Lisinopril',
         'mesh_id': 'D002316',
         'drugbank_id': 'DB00722'},
        0,
        0),
       ({'name': 'Atorvastatin',
         'synonyms': {'Atorvastatin',
          'Lipitor',
          'Liptonorm',
          'Sortis',
          'atorvastatina',
          'atorvastatine'
          'atorvastatinum'},
         'medline_plus_id': 'a600045',
         'nhs_url': 'https://www.nhs.uk/medicines/atorvastatin',
         'wikipedia_url': 'https://en.wikipedia.org/wiki/Atorvastatin',
         'mesh_id': 'D019161',
         'drugbank id': 'DB01076'},
        3,
        3)]]
[6]: w_normalized_medications = []
     for doc in w_docs:
         if doc.get("medications") is not None:
             res = []
             res = doc.get("medications").split(" ")
                 res.remove("other")
             except ValueError:
```

```
pass
    res = find_drugs(res, is_ignore_case=True)
    w_normalized_medications.append(res)
len(w_normalized_medications)

[6]: 4941
[7]: # For each nations, name out the medications and normalize them. De-dum them.
```

```
[7]: # For each patient, parse out the medications and normalize them. De-dup them,
     so each patient has each medication listed only once.
     b just names = list(
         map(lambda n: set(list(map(lambda m: m[0].get("name"), n))),__
     →b_normalized_medications)
     b_normalized_medications_names = [
         element for sublist in b just names for element in sublist
     ]
     w_just_names = list(
         map(lambda n: set(list(map(lambda m: m[0].get("name"), n))),__
      →w_normalized_medications)
     w_normalized_medications_names = [
         element for sublist in w_just_names for element in sublist
     b_just_names[:5]
     # print(len(b_normalized_medications_names))
     # print(len(w normalized medications names))
```

```
[8]: b_cv = CountVectorizer(analyzer="word")
b_cv_fit = b_cv.fit_transform(b_normalized_medications_names)
b_word_list = b_cv.get_feature_names_out()
b_count_list = b_cv_fit.toarray().sum(axis=0)

b_word_freq = dict(zip(b_word_list, b_count_list))

w_cv = CountVectorizer(analyzer="word")
w_cv_fit = w_cv.fit_transform(w_normalized_medications_names)
w_word_list = w_cv.get_feature_names_out()
w_count_list = w_cv_fit.toarray().sum(axis=0)

w_word_freq = dict(zip(w_word_list, w_count_list))
```

```
[9]: b_word_freq_df = pd.DataFrame(
          b word_freq.items(), columns=["word", "b.frequency"]
      ).sort_values(by="b.frequency", ascending=False)
      w_word_freq_df = pd.DataFrame(
          w_word_freq.items(), columns=["word", "w.frequency"]
      ).sort_values(by="w.frequency", ascending=False)
[10]: wf_df = w_word_freq_df.merge(b_word_freq_df, how="inner", on="word")
[11]: wf_df["w.frequency_pct"] = wf_df["w.frequency"] / wf_df["w.frequency"].sum()
      wf_df["b.frequency_pct"] = wf_df["b.frequency"] / wf_df["b.frequency"].sum()
      wf_df["frequency_pct_diff"] = wf_df["b.frequency_pct"] - wf_df["w.

¬frequency_pct"]

      wf_df["frequency_pct_diff_abs"] = wf_df["frequency_pct_diff"].abs()
      # Sort by largest values in absolue difference
      wf_df.sort_values(by="frequency_pct_diff", ascending=False).head(10)
                   word w.frequency b.frequency w.frequency_pct b.frequency_pct \
[11]:
              metformin
                                  201
                                               326
                                                           0.040622
                                                                             0.065912
      10
             salbutamol
                                  44
                                                62
                                                           0.008892
                                                                             0.012535
                                                62
      11
              albuterol
                                  44
                                                           0.008892
                                                                             0.012535
      8
             metoprolol
                                  87
                                               100
                                                           0.017583
                                                                             0.020218
      2
                                  343
                                               351
             amlodipine
                                                           0.069321
                                                                             0.070966
      9
               losartan
                                  50
                                                57
                                                           0.010105
                                                                             0.011524
      13 levothyroxine
                                  31
                                                37
                                                           0.006265
                                                                             0.007481
          acetaminophen
                                  286
                                               291
                                                           0.057801
                                                                             0.058835
      7
             omeprazole
                                  89
                                                93
                                                           0.017987
                                                                             0.018803
      15
               naproxen
                                   13
                                                17
                                                           0.002627
                                                                             0.003437
          frequency_pct_diff
                              frequency_pct_diff_abs
      6
                    0.025289
                                             0.025289
      10
                    0.003643
                                             0.003643
                    0.003643
      11
                                             0.003643
      8
                    0.002635
                                             0.002635
      2
                    0.001645
                                             0.001645
      9
                    0.001419
                                             0.001419
      13
                    0.001216
                                             0.001216
      4
                    0.001034
                                             0.001034
      7
                                             0.000816
                    0.000816
      15
                    0.000810
                                             0.000810
[12]: # First order frequencies by magnature of difference (absolute value), take the
       →top 200 words with the greatest difference,
      # then re-sort by actual difference so when we plot the values will be ...
       ⇒sequential from smallest to largest bars
      most = (
          wf_df.sort_values(by="frequency_pct_diff_abs", ascending=False)
```

```
chart_data = {}
      # Create a map with the word as the frequency, and the magnitude vector as the
      yalue
      # a vector of [0, n] will plot a blue bar
      # a vector of [n, 0] will plot an orange bar
      # a vector with a negative n [-n, 0] will plot a bar on the left
      # a vector with a positive n [n, 0] will plot a bar on the right
      # \{"word": [-1, 0]\} will plot an orange bar for "word" on the left of 0 with
       ⇔length 1
      # {"word": [0, 0.5]} will plot a blue bar for "word" on the right of 0 with
       ⇔length 0.5
      # in order to generate a good Positive Negative bar chart, we assign b freq to \Box
       → the left side (negative)
      # and w freq to the right side (positive)
      for row in most.iterrows():
          if row[1]["w.frequency_pct"] > row[1]["b.frequency_pct"]:
              # orange bars
              chart_data[row[1]["word"]] = [
                  row[1]["w.frequency_pct"] - row[1]["b.frequency_pct"],
              1
          else:
              # blue bars
              chart_data[row[1]["word"]] = [
                  -(row[1]["b.frequency_pct"] - row[1]["w.frequency_pct"]),
              ]
[13]: # Positive Negative Bar Chart to better visualize where word frequencies.
      ⇒diverge between data sets
      # Based on https://stackoverflow.com/a/69976552/11407943
      import numpy as np
      import matplotlib.pyplot as plt
      category_names = ["white-or-caucasian", "black-or-african-american"]
      results = chart data
      def survey(results, category_names):
          Parameters
```

.sort_values(by="frequency_pct_diff", ascending=False)

.head(200)

)

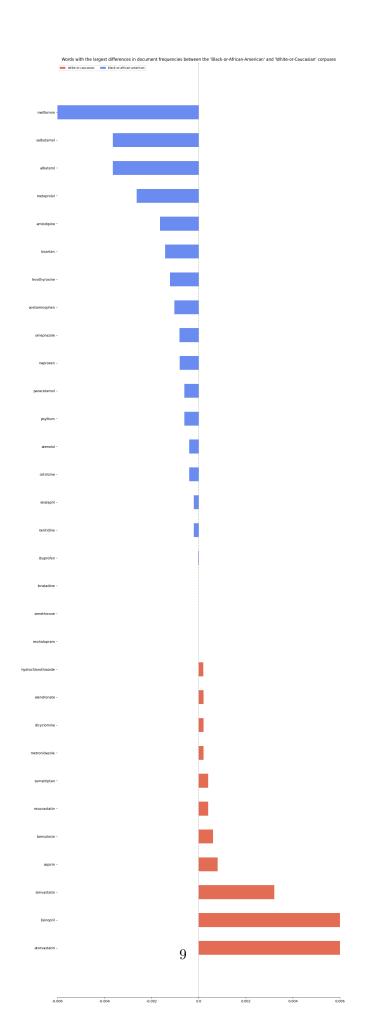
```
results : dict
      A mapping from question labels to a list of answers per category.
      It is assumed all lists contain the same number of entries and that
      it matches the length of *category_names*. The order is assumed
      to be from 'Strongly disagree' to 'Strongly aisagree'
  category_names : list of str
      The category labels.
  labels = list(results.keys())
  data = np.array(list(results.values()))
  data_cum = data.cumsum(axis=1)
  middle_index = data.shape[1] // 2
  offsets = 0 # data[:, range(middle index)].sum(axis=1) # + data[:, ____
\rightarrow middle_index]/2
  # Color Mapping
  category_colors = plt.get_cmap("coolwarm_r")(np.linspace(0.15, 0.85, data.
⇔shape[1]))
  fig, ax = plt.subplots(figsize=(15, 50))
  # Plot Bars
  for i, (colname, color) in enumerate(zip(category_names, category_colors)):
      widths = data[:, i]
      starts = data cum[:, i] - widths - offsets
      rects = ax.barh(
          labels, widths, left=starts, height=0.5, label=colname, color=color
  # Add Zero Reference Line
  ax.axvline(0, linestyle="--", color="black", alpha=0.25)
  # X Axis
  ax.set_xlim(-0.006, 0.006)
  # ax.set_xticks(np.arange(-0.0035, 0.0035, 0.003))
  ax.xaxis.set_major_formatter(lambda x, pos: str(x))
  # Y Axis
  ax.invert_yaxis()
  # Remove spines
  ax.spines["right"].set_visible(False)
  ax.spines["top"].set_visible(False)
  ax.spines["left"].set_visible(False)
```

```
# Ledgend
ax.legend(
    ncol=len(category_names),
    bbox_to_anchor=(0, 0.99),
    loc="lower left",
    fontsize="small",
)

# Set Background Color
fig.set_facecolor("#FFFFFF")

return fig, ax

fig, ax = survey(results, category_names)
plt.title(
    "Words with the largest differences in document frequencies between the_u
    'Black-or-African-American' and 'White-or-Caucasian' corpuses"
)
plt.show()
```



```
[14]: import scipy
      from sklearn.feature_extraction import text
      from collections import Counter
[15]: b_just_names_lower = [list(map(lambda x: x.lower(), arr)) for arr in__
      →b_just_names]
      b_list_of_doc_counter = list(map(Counter, b_just_names_lower))
      # element for sublist in w_just_names for element in sublist
      w_just_names_lower = [list(map(lambda x: x.lower(), arr)) for arr in__
       →w_just_names]
      w_list_of_doc_counter = list(map(Counter, w_just_names_lower))
      b just names lower
      b_medications_names_counter = Counter(
          [element for sublist in b_just_names_lower for element in sublist]
      w_medications_names_counter = Counter(
          [element for sublist in w_just_names_lower for element in sublist]
[16]: b_medications_names_counter
[16]: Counter({'lisinopril': 1974,
               'atorvastatin': 921,
               'amlodipine': 351,
               'metformin': 326,
               'ibuprofen': 301,
               'acetaminophen': 291,
               'simvastatin': 236,
               'metoprolol': 100,
               'omeprazole': 93,
               'salbutamol': 62,
               'albuterol': 62,
               'losartan': 57,
               'hydrochlorothiazide': 39,
               'levothyroxine': 37,
               'loratadine': 19,
               'naproxen': 17,
               'paracetamol': 9,
               'sumatriptan': 8,
               'aspirin': 8,
               'atenolol': 7,
               'alendronate': 6,
               'cetirizine': 5,
               'psyllium': 4,
```

```
'tamsulosin': 3,
                'pantoprazole': 2,
                'simethicone': 2,
                'celecoxib': 2,
                'glipizide': 2,
                'ranitidine': 2,
                'timolol': 2,
                'enalapril': 2,
                'gabapentin': 1,
                'serotonin': 1,
                'metronidazole': 1,
                'ciprofloxacin': 1,
                'montelukast': 1,
                'rosuvastatin': 1,
                'calcitriol': 1,
                'escitalopram': 1,
                'acenocoumarol': 1,
                'ursodiol': 1,
                'lorazepam': 1,
                'dicyclomine': 1,
                'amitriptyline': 1,
                'bisacodyl': 1})
[17]: w_medications_names_counter
[17]: Counter({'lisinopril': 2031,
               'atorvastatin': 1055,
               'amlodipine': 343,
                'ibuprofen': 301,
                'acetaminophen': 286,
                'simvastatin': 252,
                'metformin': 201,
                'omeprazole': 89,
                'metoprolol': 87,
                'losartan': 50,
                'salbutamol': 44,
                'albuterol': 44,
                'hydrochlorothiazide': 40,
                'levothyroxine': 31,
                'loratadine': 19,
               'naproxen': 13,
                'aspirin': 12,
                'sumatriptan': 10,
                'alendronate': 7,
                'paracetamol': 6,
                'tamsulosin': 6,
                'sertraline': 5,
```

```
'atenolol': 5,
               'cetirizine': 3,
               'rosuvastatin': 3,
               'simethicone': 2,
               'oxycodone': 2,
               'famotidine': 2,
               'metronidazole': 2,
               'docusate': 2,
               'dicyclomine': 2,
               'meloxicam': 1,
               'ranitidine': 1.
               'benazepril': 1,
               'furosemide': 1,
               'escitalopram': 1,
               'enalapril': 1,
               'methimazole': 1,
               'esomeprazole': 1,
               'lansoprazole': 1,
               'fluoxetine': 1,
               'psyllium': 1})
[18]: total keys = list(
          set(
              list(w_medications_names_counter.keys())
              + list(b_medications_names_counter.keys())
          )
      )
      new_counts = {}
      aa = []
      ca = []
      for k in total_keys:
          # [aa,ca]
          new counts[k] = [
              b_medications_names_counter.get(k, 0),
              w_medications_names_counter.get(k, 0),
          aa.append(b medications names counter.get(k, 0))
          ca.append(w_medications_names_counter.get(k, 0))
      c_table = pd.DataFrame.from_dict(new_counts)
      c_table.rename(index={0: "b.freq"}, inplace=True)
      c_table.rename(index={1: "w.freq"}, inplace=True)
      c_table
[18]:
              aspirin serotonin furosemide ursodiol famotidine metoprolol \
                    8
                                                                             100
      b.freq
                               1
                                            0
                                                      1
                   12
                               0
                                            1
                                                      0
                                                                  2
                                                                              87
      w.freq
```

```
b.freq
                     0
                                     1
                                                   291
                                                                 0 ...
                     2
                                     0
                                                   286
                                                                 5 ...
                                                                              0
      w.freq
              alendronate atorvastatin calcitriol sumatriptan fluoxetine
                                     921
                                                    1
      b.freq
                        6
                                    1055
                        7
                                                    0
                                                                              1
      w.freq
                                                                10
              atenolol metformin oxycodone
                                               omeprazole
                               326
      b.freq
                     7
                                                        93
      w.freq
                     5
                               201
                                            2
                                                        89
      [2 rows x 56 columns]
[19]: class bcolors:
          HEADER = "\033[95m"]
          OKBLUE = "\033[94m"]
          OKCYAN = "\033[96m"
          OKGREEN = "\033[92m"]
          WARNING = \sqrt{033}[93m]
          FAIL = "\033[91m"]
          ENDC = "\033[0m"]
          BOLD = "\033[1m"]
          UNDERLINE = \sqrt{033[4m]}
[20]: sig_results = []
      # Chi square independence test
      # https://www.dir.uniupo.it/pluqinfile.php/138296/mod resource/content/0/
       →22-colloc-bw.pdf
      for k in list(set(total keys)):
          # For AA [Number of instances of current word, Number of instances of all_
       ⇔other words]
          x1 = [c_table[k].iloc[0], c_table.iloc[0].sum() - c_table[k].iloc[0]]
          # For CA [Number of instances of current word, Number of instances of all_{\sqcup}
       ⇔other words]
          y1 = [c_table[k].iloc[1], c_table.iloc[1].sum() - c_table[k].iloc[1]]
          test = scipy.stats.chi2_contingency([x1, y1])
          word = c_table[k].name
          if test.pvalue < 0.05:</pre>
              sig_results.append(word)
              print(f"{bcolors.OKGREEN}{bcolors.BOLD}Medication: {k}{bcolors.ENDC}")
              print(f"AA: {x1}")
              print(f"CA: {y1}")
              print(
```

docusate acenocoumarol acetaminophen sertraline ... timolol \

```
f'There {bcolors.OKGREEN}is a significant difference{bcolors.ENDC}

in the frequency of the word {word} with a p-value of {bcolors.OKGREEN +"{:0.

if '.format(test.pvalue) + bcolors.ENDC}'

print(f"")

if len(sig_results) == 0:

print(f'{bcolors.BOLD}{bcolors.FAIL}No significant differences in any

conditions between groups found{bcolors.ENDC}')
```

Medication: atorvastatin

AA: [921, 4043] CA: [1055, 3911]

There is a significant difference in the frequency of the word atorva statin with a p-value of 0.001

Medication: metformin

AA: [326, 4638] CA: [201, 4765]

There is a significant difference in the frequency of the word metformin with a p-value of 0.000