

## 5\_pmh\_analysis

August 1, 2023

```
[1]: import json
import os
from sklearn.feature_extraction.text import CountVectorizer
import pandas as pd
import numpy as np
import json5
import spacy
import medspacy
import random
from medspacy.ner import TargetMatcher, TargetRule
from medspacy.visualization import visualize_ent, visualize_dep
from spacy.tokens import Span
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 = "shortness-of-breath"
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 / "white-or-caucasian"
)
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))

```

4933

4935

```

[4]: # Grab the text from each document's past medical history section
     b_pmh = []

```

```

for doc in b_docs:
    if doc.get("past_medical_history") is not None:
        b_pmh.append(doc.get("past_medical_history"))

w_pmh = []
for doc in w_docs:
    if doc.get("past_medical_history") is not None:
        w_pmh.append(doc.get("past_medical_history"))

```

```

[5]: # We want to see if each patient has a history of any of the following
      ↪ conditions
nlp = medspacy.load()
print(nlp.pipe_names)

try:
    Span.set_extension("icd10_code", default="")
except:
    pass

# Add rules for target concept extraction
target_matcher = nlp.get_pipe("medspacy_target_matcher")
target_rules = [
    TargetRule("hypertension", category="CONDITION", attributes={"icd10_code":
        ↪ "I10"}),
    TargetRule(
        "hyperlipidemia", category="CONDITION", attributes={"icd10_code": "E78.
        ↪ 5"}
    ),
    TargetRule(
        "osteoarthritis", category="CONDITION", attributes={"icd10_code": "M19.
        ↪ 90"}
    ),
    TargetRule(
        "osteoporosis", category="CONDITION", attributes={"icd10_code": "M81.0"}
    ),
    TargetRule(
        "dyslipidemia", category="CONDITION", attributes={"icd10_code": "E78.5"}
    ),
    TargetRule(
        literal="Type II Diabetes Mellitus",
        category="CONDITION",
        attributes={"icd10_code": "E11.9"},
    ),
    TargetRule(
        literal="diabetes mellitus type 2",
        category="CONDITION",
        pattern=[

```

```

        {"LOWER": "diabetes"},
        {"LOWER": "mellitus"},
        {"LOWER": "type"},
        {"LOWER": {"IN": ["two", "ii", "2"]}},
    ],
    attributes={"icd10_code": "E11.9"},
),
TargetRule(
    literal="gerd",
    category="CONDITION",
    pattern=[
        {"LOWER": "gastroesophageal"},
        {"LOWER": "reflux"},
        {"LOWER": "disease"},
    ],
    attributes={"icd10_code": "K21.9"},
),
TargetRule(
    literal="GERD", category="CONDITION", attributes={"icd10_code": "K21.9"}
),
TargetRule(
    literal="Type II Diabetes Mellitus",
    category="CONDITION",
    pattern=[
        {"LOWER": "type"},
        {"LOWER": {"IN": ["two", "ii", "2"]}},
        {
            "LOWER": {
                "IN": [
                    "dm",
                    "diabetes mellitus",
                    "diabetes",
                ]
            }
        },
    ],
    attributes={"icd10_code": "E11.9"},
),
TargetRule("asthma", category="CONDITION", attributes={"icd10_code": "J45"}),
TargetRule(
    "atrial fibrillation",
    category="CONDITION",
    attributes={"icd10_code": "I48.91"},
),
TargetRule(
    "hypercholesterolemia",

```

```

        category="CONDITION",
        attributes={"icd10_code": "E78.00"},
    ),
    TargetRule(
        "high cholesterol",
        category="CONDITION",
        pattern=[{"LOWER": {"IN": ["high", "elevated"]}}, {"LOWER": "
↪cholesterol"}],
        attributes={"icd10_code": "E78.00"},
    ),
    TargetRule(
        "hypertriglyceridemia", category="CONDITION", attributes={"icd10_code": "
↪E78.1"}
    ),
    TargetRule(
        "myocardial infarction",
        category="CONDITION",
        pattern=[
            {"LOWER": "myocardial"},
            {"LOWER": "infarction"},
        ],
        attributes={"icd10_code": "I21.9"},
    ),
    TargetRule(
        "coronary artery disease",
        category="CONDITION",
        attributes={"icd10_code": "I25.10"},
    ),
    TargetRule(
        "Irritable Bowel Syndrome",
        category="CONDITION",
        pattern=[
            {"LOWER": "irritable"},
            {"LOWER": "bowel"},
            {"LOWER": "syndrome"},
        ],
        attributes={"icd10_code": "K58"},
    ),
    TargetRule(
        "IBS",
        category="CONDITION",
        pattern=[
            {"LOWER": "ibs"},
        ],
        attributes={"icd10_code": "K58"},
    ),
    TargetRule(

```

```

        "Nephrolithiasis",
        category="CONDITION",
        pattern=[
            {"LOWER": "nephrolithiasis"},
        ],
        attributes={"icd10_code": "N20.0"},
    ),
    TargetRule(
        "Kidney Stones",
        category="CONDITION",
        pattern=[
            {"LOWER": "kidney"},
            {
                "LOWER": {
                    "IN": [
                        "stones",
                        "stone",
                    ]
                },
            },
        ],
        attributes={"icd10_code": "N20.0"},
    ),
    TargetRule(
        "Gallstones",
        category="CONDITION",
        pattern=[
            {"LOWER": "gallstones"},
        ],
        attributes={"icd10_code": "K80"},
    ),
    TargetRule(
        "Cholelithiasis",
        category="CONDITION",
        pattern=[
            {"LOWER": "cholelithiasis"},
        ],
        attributes={"icd10_code": "K80"},
    ),
    TargetRule(
        "Diverticulosis",
        category="CONDITION",
        pattern=[
            {"LOWER": "diverticulosis"},
        ],
        attributes={"icd10_code": "K57.9"},
    ),

```

```

TargetRule(
    "Endometriosis",
    category="CONDITION",
    pattern=[
        {"LOWER": "endometriosis"},
    ],
    attributes={"icd10_code": "N80.9"},
),
TargetRule(
    "Appendicitis",
    category="CONDITION",
    pattern=[
        {"LOWER": "appendicitis"},
    ],
    attributes={"icd10_code": "K35.80"},
),
TargetRule(
    "Migraine",
    category="CONDITION",
    pattern=[
        {"LOWER": "migraine"},
        {"LOWER": "migraines"},
    ],
    attributes={"icd10_code": "G43.909"},
),
TargetRule(
    "Pancreatitis",
    category="CONDITION",
    pattern=[
        {"LOWER": "pancreatitis"},
    ],
    attributes={"icd10_code": "K85.9"},
),
TargetRule(
    "Cholecystitis",
    category="CONDITION",
    pattern=[
        {"LOWER": "cholecystitis"},
    ],
    attributes={"icd10_code": "K81"},
),
TargetRule(
    "Diverticulitis",
    category="CONDITION",
    pattern=[
        {"LOWER": "diverticulitis"},
    ],

```

```

        attributes={"icd10_code": "K57.92"},
    ),
    TargetRule(
        "Gastritis",
        category="CONDITION",
        pattern=[
            {"LOWER": "gastritis"},
        ],
        attributes={"icd10_code": "K29"},
    ),
    TargetRule(
        "Gastric Ulcers",
        category="CONDITION",
        pattern=[
            {"LOWER": "gastric"},
            {"LOWER": {"IN": ["ulcers", "ulcer"]}},
        ],
        attributes={"icd10_code": "K25.9"},
    ),
    TargetRule(
        "Constipation",
        category="CONDITION",
        pattern=[
            {"LOWER": "constipation"},
        ],
        attributes={"icd10_code": "K59.00"},
    ),
    TargetRule(
        "COPD",
        category="CONDITION",
        pattern=[
            {"LOWER": "copd"},
        ],
        attributes={"icd10_code": "J44.9"},
    ),
    TargetRule(
        "Chronic Obstructive Pulmonary Disease",
        category="CONDITION",
        pattern=[
            {"LOWER": "chronic"},
            {"LOWER": "obstructive"},
            {"LOWER": "pulmonary"},
            {"LOWER": "disease"},
        ],
        attributes={"icd10_code": "J44.9"},
    ),
    TargetRule(

```



```

        "Other Seasonal Allergic Rhinitis",
        category="CONDITION",
        attributes={"icd10_code": "J30.2"},
    ),
    TargetRule(
        "Seasonal Allergies",
        category="CONDITION",
        pattern=[
            {"LOWER": "seasonal"},
            {"LOWER": {"IN": ["allergies", "allergy"]}},
        ],
        attributes={"icd10_code": "J30.2"},
    ),
    TargetRule(
        "Congestive Heart Failure",
        category="CONDITION",
        pattern=[
            {"LOWER": "congestive"},
            {"LOWER": "heart"},
            {"LOWER": "failure"},
        ],
        attributes={"icd10_code": "I50.9"},
    ),
    TargetRule(
        "CHF",
        category="CONDITION",
        pattern=[
            {"LOWER": "chf"},
        ],
        attributes={"icd10_code": "I50.9"},
    ),
    TargetRule(
        "Hypothyroidism",
        category="CONDITION",
        pattern=[
            {"LOWER": "hypothyroidism"},
        ],
        attributes={"icd10_code": "E03.9"},
    ),
    TargetRule(
        "Hypothyroid",
        category="CONDITION",
        pattern=[
            {"LOWER": "hypothyroid"},
        ],
        attributes={"icd10_code": "E03.9"},
    ),

```

```

TargetRule(
    "Hyperthyroidism",
    category="CONDITION",
    pattern=[
        {"LOWER": "hyperthyroidism"},
    ],
    attributes={"icd10_code": "E05"},
),
TargetRule(
    "Hyperthyroid",
    category="CONDITION",
    pattern=[
        {"LOWER": "hyperthyroid"},
    ],
    attributes={"icd10_code": "E05"},
),
TargetRule(
    "High Blood Pressure",
    category="CONDITION",
    pattern=[
        {"LOWER": "high"},
        {"LOWER": "blood"},
        {"LOWER": "pressure"},
    ],
    attributes={"icd10_code": "I10"},
),
]

ICD_TO_TEXT_MAP = {
    "I10": "hypertension",
    "E78.5": "hyperlipidemia",
    "M19.90": "osteoarthritis",
    "E11.9": "type ii diabetes mellitus",
    "E78.00": "hypercholesterolemia",
    "J45": "asthma",
    "I48.91": "atrial fibrillation",
    "M81.0": "osteoporosis",
    "K21.9": "gastroesophageal reflux disease ",
    "I21.9": "myocardial infarction",
    "I25.10": "coronary artery disease",
    "K85.9": "pancreatitis",
    "G43.909": "migraine",
    "K35.80": "appendicitis",
    "N80.9": "endometriosis",
    "K57.9": "diverticulosis",
    "K80": "cholelithiasis",

```

```

    "N20.0": "nephrolithiasis",
    "K58": "ibs",
    "K81": "cholecystitis",
    "K57.92": "diverticulitis",
    "K29": "gastritis",
    "K25.9": "gastric ulcers",
    "K59.00": "constipation",
    "J44.9": "copd",
    "J30.2": "other seasonal allergic rhinitis",
    "I50.9": "congestive heart failure",
    "E03.9": "hypothyroidism",
    "E05": "hyperthyroidism",
}

```

```
target_matcher.add(target_rules)
```

```
['medspacy_pyrush', 'medspacy_target_matcher', 'medspacy_context']
```

```
[6]: # Extract conditions from PMH
```

```

b_nlp_pmh = []
for doc in b_pmh:
    doc = nlp(doc)
    b_nlp_pmh.append(doc)

```

```
[7]: w_nlp_pmh = []
```

```

for doc in w_pmh:
    doc = nlp(doc)
    w_nlp_pmh.append(doc)

```

```
[8]: # Quick test to make sure negation detection works
```

```
# negation test
```

```

test = "The patient has a history of hypertension which is well-controlled with
↳medication. She also has a history of gallstones but has not had any
↳previous episodes of cholecystitis or pancreatitis ."

```

```

doc = nlp(test)
visualize_ent(doc)
for ent in doc.ents:
    print(ent._.is_negated)

```

```
<IPython.core.display.HTML object>
```

```
False
```

```
False
```

```
True
```

```
True
```

```
[9]: # Quick visualization of entity extraction
```

```

for doc in w_nlp_pmh[:1000]:
    visualize_ent(doc)

```



[illegible]

[illegible]



[illegible]



[illegible]

[illegible]

[illegible]



[illegible]

[illegible]

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[illegible]





[illegible]

[illegible]

```

<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>

```

```

[10]: for doc in b_nlp_pmh[:10]:
        visualize_ent(doc)

```

```

<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>
<IPython.core.display.HTML object>

```

```

[11]: # Test entity extraction, making sure to ignore negated entities
test = b_nlp_pmh[0:2]
test.append(
    nlp(
        "patient admits to type 2 diabetes but denies any hypertension. he
        ↪takes metformin for his type 2 diabetes."
    )
)
print(list(map(lambda x: [y for y in list(x.ents) if y._.is_negated == False],
    ↪test)))
print(
    list(
        map(
            lambda x: [y._.icd10_code for y in list(x.ents) if y._.is_negated
            ↪== False],
            test,
        )
    )
)

```

```

    )
)
# De-dup
print(
    list(
        map(
            lambda x: set(
                [y._icd10_code for y in list(x.ents) if y._is_negated ==
↪False]
            ),
            test,
        )
    )
)

```

```

[[hypertension, hyperlipidemia, coronary artery disease], [Hypertension,
hyperlipidemia], [type 2 diabetes, type 2 diabetes]]
[['I10', 'E78.5', 'I25.10'], ['I10', 'E78.5'], ['E11.9', 'E11.9']]
[{'I10', 'E78.5', 'I25.10'}, {'I10', 'E78.5'}, {'E11.9'}]

```

```

[12]: # Do entity extraction on the PMH section of the notes, skipping negated
↪entities. Make sure to de-duplicate the entities.
b_just_names = list(
    map(
        lambda x: set(
            [y._icd10_code for y in list(x.ents) if y._is_negated == False]
        ),
        b_nlp_pmh,
    )
)

b_normalized_conditions_names = [
    element for sublist in b_just_names for element in sublist
]

w_just_names = list(
    map(
        lambda x: set(
            [y._icd10_code for y in list(x.ents) if y._is_negated == False]
        ),
        w_nlp_pmh,
    )
)

w_normalized_conditions_names = [
    element for sublist in w_just_names for element in sublist
]

print(len(b_normalized_conditions_names))
print(len(w_normalized_conditions_names))

```

6179

6223

```
[13]: # Count the instances of each word in the black and white conditions.
      ↪ Conditions are de-duped, so if a condition appears multiple times in a
      ↪ single participant's data, it is only counted once.
      # We fix this later before doing statistical analysis.
      from collections import Counter

      b_word_freq = Counter(b_normalized_conditions_names)
      w_word_freq = Counter(w_normalized_conditions_names)
```

```
[14]: 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)
```

```
[15]: wf_df = w_word_freq_df.merge(b_word_freq_df, how="inner", on="word")
      wf_df
```

```
[15]:
```

	word	w.frequency	b.frequency
0	I10	2438	2514
1	E78.5	1499	1444
2	J45	1011	994
3	J44.9	487	409
4	M19.90	278	231
5	J30.2	201	204
6	I50.9	97	117
7	E11.9	82	124
8	I25.10	44	49
9	E78.00	21	23
10	M81.0	18	11
11	I21.9	16	27
12	E03.9	15	12
13	I48.91	13	14
14	K21.9	2	5

```
[16]: 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 absolute difference
      wf_df.sort_values(by="frequency_pct_diff", ascending=False).head(25)
```

```
[16]:
```

	word	w.frequency	b.frequency	w.frequency_pct	b.frequency_pct	\
0	I10	2438	2514	0.391835	0.406928	
7	E11.9	82	124	0.013179	0.020071	
6	I50.9	97	117	0.015590	0.018938	
11	I21.9	16	27	0.002572	0.004370	
8	I25.10	44	49	0.007072	0.007931	
5	J30.2	201	204	0.032305	0.033020	
14	K21.9	2	5	0.000321	0.000809	
9	E78.00	21	23	0.003375	0.003723	
13	I48.91	13	14	0.002089	0.002266	
12	E03.9	15	12	0.002411	0.001942	
10	M81.0	18	11	0.002893	0.001781	
2	J45	1011	994	0.162488	0.160893	
1	E78.5	1499	1444	0.240919	0.233733	
4	M19.90	278	231	0.044680	0.037391	
3	J44.9	487	409	0.078271	0.066203	

	frequency_pct_diff	frequency_pct_diff_abs
0	0.015092	0.015092
7	0.006892	0.006892
6	0.003348	0.003348
11	0.001799	0.001799
8	0.000860	0.000860
5	0.000716	0.000716
14	0.000488	0.000488
9	0.000348	0.000348
13	0.000177	0.000177
12	-0.000468	0.000468
10	-0.001112	0.001112
2	-0.001594	0.001594
1	-0.007187	0.007187
4	-0.007289	0.007289
3	-0.012068	0.012068

```
[17]: # First order frequencies by magnitude 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)
    .head(200)
    .sort_values(by="frequency_pct_diff", ascending=False)
)

chart_data = {}
```

```

# Create a map with the word as the frequency, and the magnitude vector as the
↪value\
# 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
↪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"],
            0,
        ]
    else:
        # blue bars
        chart_data[row[1]["word"]] = [
            0,
            -(row[1]["b.frequency_pct"] - row[1]["w.frequency_pct"]),
        ]

```

```

[18]: # 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
    -----
    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 agree'
    """

```

```

category_names : list of str
    The category labels.
"""

labels = list(map(lambda i: ICD_TO_TEXT_MAP.get(i), 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[:,
↪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)

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

```



```

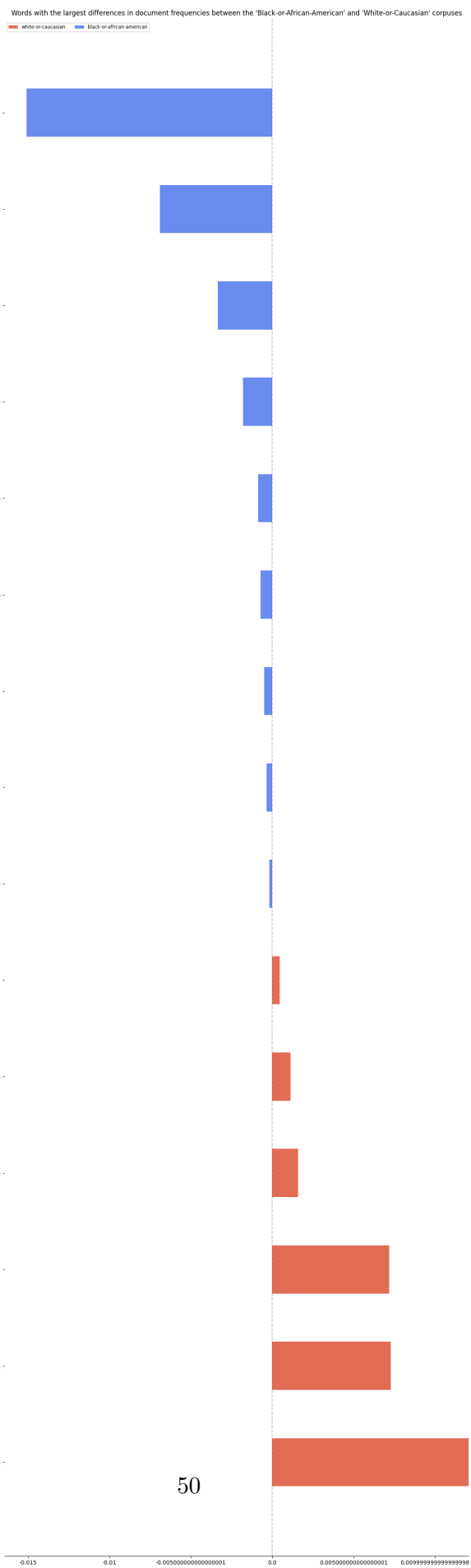
)

# 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
    ↪ 'Black-or-African-American' and 'White-or-Caucasian' corpuses"
)
plt.show()

```



```
[19]: import scipy
      from sklearn.feature_extraction import text
      from collections import Counter
```

```
[20]: b_just_names_lower = [
      list(map(lambda x: ICD_TO_TEXT_MAP.get(x), 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: ICD_TO_TEXT_MAP.get(x), arr)) for arr in w_just_names
      ]
      w_list_of_doc_counter = list(map(Counter, w_just_names_lower))
      b_conditions_names_counter = Counter(
      [element for sublist in b_just_names_lower for element in sublist]
      )
      w_conditions_names_counter = Counter(
      [element for sublist in w_just_names_lower for element in sublist]
      )
```

```
[21]: b_conditions_names_counter
```

```
[21]: Counter({'hypertension': 2514,
              'hyperlipidemia': 1444,
              'asthma': 994,
              'copd': 409,
              'osteoarthritis': 231,
              'other seasonal allergic rhinitis': 204,
              'type ii diabetes mellitus': 124,
              'congestive heart failure': 117,
              'coronary artery disease': 49,
              'myocardial infarction': 27,
              'hypercholesterolemia': 23,
              'atrial fibrillation': 14,
              'hypothyroidism': 12,
              'osteoporosis': 11,
              'gastroesophageal reflux disease ': 5,
              'constipation': 1})
```

```
[22]: w_conditions_names_counter
```

```
[22]: Counter({'hypertension': 2438,
              'hyperlipidemia': 1499,
              'asthma': 1011,
              'copd': 487,
```

```

'osteoarthritis': 278,
'other seasonal allergic rhinitis': 201,
'congestive heart failure': 97,
'type ii diabetes mellitus': 82,
'coronary artery disease': 44,
'hypercholesterolemia': 21,
'osteoporosis': 18,
'myocardial infarction': 16,
'hypothyroidism': 15,
'atrial fibrillation': 13,
'gastroesophageal reflux disease ': 2,
'cholelithiasis': 1})

```

```

[23]: total_keys = list(
    set(
        list(w_conditions_names_counter.keys())
        + list(b_conditions_names_counter.keys())
    )
)
new_counts = {}
aa = []
ca = []
for k in total_keys:
    # [aa,ca]
    new_counts[k] = [
        b_conditions_names_counter.get(k, 0),
        w_conditions_names_counter.get(k, 0),
    ]
    aa.append(b_conditions_names_counter.get(k, 0))
    ca.append(w_conditions_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

```

```

[23]:      asthma  atrial fibrillation  osteoarthritis  myocardial infarction  \
b.freq      994                14             231                27
w.freq     1011                13             278                16

      hypercholesterolemia  congestive heart failure  copd  hypothyroidism  \
b.freq                  23                117    409                12
w.freq                  21                97    487                15

      other seasonal allergic rhinitis  constipation  osteoporosis  \
b.freq                  204                1            11
w.freq                  201                0            18

```

	type ii diabetes mellitus	hyperlipidemia	hypertension	\
b.freq	124	1444	2514	
w.freq	82	1499	2438	

	coronary artery disease	gastroesophageal reflux disease	\
b.freq	49	5	
w.freq	44	2	

	cholelithiasis
b.freq	0
w.freq	1

```
[24]: class bcolors:
    HEADER = "\033[95m"
    OKBLUE = "\033[94m"
    OKCYAN = "\033[96m"
    OKGREEN = "\033[92m"
    WARNING = "\033[93m"
    FAIL = "\033[91m"
    ENDC = "\033[0m"
    BOLD = "\033[1m"
    UNDERLINE = "\033[4m"

[25]: sig_results = []
# Chi square independence test
# https://www.dir.uniupo.it/pluginfile.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
    ↳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:
        sig_results.append(word)
        print(f"{bcolors.BOLD}Condition: {k}{bcolors.ENDC}")
        print(f"    W    ^W")
        print(f"AA: {x1}")
        print(f"CA: {y1}")
        print(
            f'There {bcolors.OKGREEN}is a significant difference{bcolors.ENDC}
↳in the prevalence of the condition "{word}" between the groups with a
↳p-value of {bcolors.OKGREEN + "{:0.3f}".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}"
    )

```

Condition: osteoarthritis

W    ^W

AA: [231, 5948]

CA: [278, 5945]

There is a significant difference in the prevalence of the condition  
 "osteoarthritis" between the groups with a p-value of 0.045

Condition: copd

W    ^W

AA: [409, 5770]

CA: [487, 5736]

There is a significant difference in the prevalence of the condition  
 "copd" between the groups with a p-value of 0.010

Condition: type ii diabetes mellitus

W    ^W

AA: [124, 6055]

CA: [82, 6141]

There is a significant difference in the prevalence of the condition  
 "type ii diabetes mellitus" between the groups with a p-value of 0.003