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October 8, 2024

0.0.1 Problem Statement:

0.0.2 This analysis aims to uncover potential causes of mental health disorders by examining the behavioral patterns and histories of affected patients, ### leading to insights for better interventions.

```
[1]: import pandas as pd
import os
from typing import List, Optional, Dict
import gc
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: # !pip3 install pandas
#!pip3 install pyarrow
# !pip3 install pandas
#!pip3 install fastparquet
```

0.0.3 Data Fetch

```
[3]: def fetch_nsduh_data(year: int) -> Optional[pd.DataFrame]:
    url_placeholder = "https://www.datafiles.samhsa.gov/sites/default/files/
    field-uploads-protected/studies/NSDUH-{year}-NSDUH-{year}-datasets/
    NSDUH-{year}-DS0001/NSDUH-{year}-DS0001-bundles-with-study-info/
    NSDUH-{year}-DS0001-bndl-data-tsv.zip"

    try:
        url = url_placeholder.format(year=year)
        df = pd.read_csv(url, compression='zip', sep='\t', low_memory=False)
        return df
    except Exception as e:
        print(f"An error occurred for year {year}: {e}")
        return None
```

```
[4]: def check_parquet_exists(years: List[int], output_dir: str) -> Dict[int, bool]:
    existence_check = {}
```

```
for year in years:
             year_path = os.path.join(output_dir, f'year={year}')
             existence_check[year] = os.path.exists(year_path)
        return existence_check
[5]: def write_parquet(df: pd.DataFrame, year: int, output_dir: str, overwrite: boolu
      ⇒= False) -> None:
        try:
            year_dir = os.path.join(output_dir, f'year={year}')
             if overwrite and os.path.exists(year_dir):
                 shutil.rmtree(year_dir) # Remove existing directory to start fresh
            df['year'] = year # Add the year column for partitioning
             # Write data to Parquet format with partitioning
            df.to_parquet(output_dir, partition_cols=['year'], index=False)
            print(f"Data for year {year} successfully saved to Parquet format in ⊔

√{output_dir}")
         except Exception as e:
            print(f"Error saving data to Parquet for year {year}: {e}")
[6]: def data_fetch(years_to_fetch: List[int], output_dir: str, overwrite: bool =__
      →False) -> None:
        try:
            for year in years_to_fetch:
                 if not overwrite and os.path.exists(os.path.join(output_dir,_
      print(f"Data for year {year} already exists. Skipping.")
                     continue
                df = fetch_nsduh_data(year)
                 if df is not None:
                    print(f"Successfully fetched data for year: {year}")
                    write_parquet(df, year, output_dir, overwrite)
                    del df # Remove the DataFrame from memory
                    gc.collect() # Force garbage collection
            print("All requested years processed.")
         except Exception as e:
            print(f"An unexpected error occurred in the data fetch function: {e}")
[7]: def read_parquet(input_dir: str, years: Optional[List[int]] = None) -> pd.
      →DataFrame:
        data frames = []
        available_years = [int(d.split('=')[1]) for d in os.listdir(input_dir) if d.
      ⇔startswith('year=')]
```

```
years_to_read = years if years is not None else available_years
for year in years_to_read:
    year_path = os.path.join(input_dir, f'year={year}')
    if os.path.exists(year_path):
        df = pd.read_parquet(year_path)
        # Convert data types to reduce memory usage
        for col in df.select dtypes(include=['float64']).columns:
            df[col] = df[col].astype('float32')
        data_frames.append(df)
    else:
        print(f"Warning: No data found for year {year}")
if data_frames:
    combined_df = pd.concat(data_frames, ignore_index=True)
    return combined_df
else:
    print("Warning: No data was loaded.")
    return pd.DataFrame()
```

```
if __name__ == "__main__":
    years = [2015, 2016, 2017, 2018, 2019]
    output_directory = "../data/DS/NSDUH"
    data_fetch(years, output_directory, overwrite=False)
    for year in years:
        df = read_parquet(output_directory, [year])
        if year in df:
            print(f"Data for year {year}:")
            print(df[year].head())
    #del df
        gc.collect()
```

Data for year 2015 already exists. Skipping. Data for year 2016 already exists. Skipping. Data for year 2017 already exists. Skipping. Data for year 2018 already exists. Skipping. Data for year 2019 already exists. Skipping. All requested years processed.

- 0.0.4 Name : Rama Rao Vydadi
- 0.0.5 Person Number: 50604256
- 0.0.6 Question 1: How does socioeconomic status (income, education, employment status) influence the likelihood of experiencing mental health disorders?
- 0.0.7 Hypothesis 1: Individuals with lower income are more likely to experience mental health issues or Unemployment is a significant predictor of mental health disorders
- 0.0.8 Significance of the question: Understanding the correlation between socioeconomic factors and mental health can help identify vulnerable populations and provide support for economically weak population.
- 0.0.9 Question2: What role does marijuana use play in the aggravation of mental health disorders?
- 0.0.10 Hypothesis 2: Frequent marijuana use is more common in states where marijuana is legalized, and this is associated with a higher prevalence of depression.
- 0.0.11 Understanding the relationship between marijuana use and mental health will help us to answer sensitive questions like should marijuana be banned all over the world
- 0.0.12 Question 3: What role does hallucing ens play in mental health disorders?
- 0.0.13 Hypothesis 3: Unregulated use of hallucinogens can cause mental health problems like anxiety and depression.
- 0.0.14 Understanding the relationship between use of various hallucinogens and mental health will suggest us to regulate the supply of medical hallucinogens

#We will first perform the basic data cleaning steps and then perform EDA

[9]:	df	.head()									
[9]:		QUESTID2 FILEDATE		CIGEVER	CIGOFRSM	CIGWILYR	CIGTRY	CIGYFU	CIGMFU	\	
	0	43295143	10/09/202	20	1	99	99	13	9999	99	
	1	65095143	10/09/202	20	2	99	99	991	9991	91	
	2	49405143	10/09/202	20	1	99	99	22	9999	99	
	3	51015143	10/09/202	20	2	99	99	991	9991	91	
	4	31825143	10/09/202	20	2	99	99	991	9991	91	
		CIGREC (CIG3OUSE		POVERTY3	TOOLONG	TROUBUND	PDEN10	COUTYP4	\	
	0	4	93		3.0	2	2	2	2		
	1	91	91		3.0	2	2	2	2		
	2	4	93		3.0	2	2	2	2		
	3	91	91		1.0	2	2	2	2		
	4	91	91		3.0	2	2	2	2		
		MAIIN102	AIIND102		ANALWT_C	VESTR	VEREP				
	0	2	2	6	613.865723	40004	2				

```
2
                   2 6321.580566 40003
1
                                             1
2
         2
                   2 5045.607422 40008
                                             1
3
         2
                   2 2419.558838 40031
                                             1
4
         2
                       575.225464 40010
                                             2
```

[5 rows x 2741 columns]

[10]: df.shape

[10]: (56136, 2741)

[11]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 56136 entries, 0 to 56135
Columns: 2741 entries, QUESTID2 to VEREP
dtypes: float32(407), int64(2332), object(2)

memory usage: 1.1+ GB

[12]: df.describe()

[12]:			QUESTID2	(CIGEVER		CIGOFRSM		CIGWILYR		CIGTRY	\
	count	5.	613600e+04	56136	.000000	561	136.000000	56	136.000000	56	136.000000	
	mean	5.	434607e+07	1	.542700		78.595554		78.605458		550.575816	
	std	2.	563167e+07	0	.498178		39.056828		39.037478		485.236660	
	min	1.	000945e+07	1	.000000		1.000000		1.000000		1.000000	
	25%	3.	198245e+07	1	.000000		99.000000		99.000000		16.000000	
	50%	5.	403939e+07	2	.000000		99.000000		99.000000		991.000000	
	75%	7.	625105e+07	2	.000000		99.000000		99.000000		991.000000	
	max	9.	999669e+07	2	.000000		99.000000		99.000000		997.000000	
			CIGYFU		CIGMFU		CIGREC		CIG30USE		CG30EST	\
	count	56	136.000000	56136	.000000	561	136.000000	56	136.000000	56	136.000000	
	mean	9	840.048846	92	.951600		50.651080		80.111248		92.784915	
	std	1	099.156156	12	. 225758		43.980321		26.734751		4.208800	
	min	2	017.000000	1	.000000		1.000000		1.000000		1.000000	
	25%	9	991.000000	91	.000000		3.000000		91.000000		91.000000	
	50%	9	991.000000	91	.000000		91.000000		91.000000		91.000000	
	75%	9	999.000000	99	.000000		91.000000		93.000000		93.000000	
	max	9	999.000000	99	.000000		91.000000		98.000000		99.000000	
		•••	POVERT		TOOLO		TROUB		PDEN	110	\	
	count	•••	55609.0000	00 56:	136.0000	00	56136.000	000	56136.0000	000		
	mean	•••	2.4271	79	2.1736	50	2.199	854	1.6629	97		
	std	•••	0.7767	59	4.7979	38	4.794	185	0.6271	.46		
	min	•••	1.0000		1.0000		1.000		1.0000			
	25%	•••	2.0000	00	2.0000		2.000		1.0000			
	50%	•••	3.0000	00	2.0000	00	2.000	000	2.0000	000		

```
75%
                    3.000000
                                   2.000000
                                                 2.000000
                                                                2.000000
                    3.000000
                                  98.000000
                                                98.000000
                                                                3.000000
      max
                  COUTYP4
                                MAIIN102
                                              AIIND102
                                                             ANALWT_C
                                                                              VESTR \
             56136.000000
                           56136.000000
                                          56136.000000 56136.000000
                                                                       56136.000000
      count
                 1.747827
                                1.982827
                                              1.982560
                                                         4902.758301
                                                                       40025.570899
      mean
                                                                          14.388113
      std
                 0.762371
                                0.129915
                                                         5952.114746
                                              0.130904
     min
                 1.000000
                                1.000000
                                              1.000000
                                                             3.581148 40001.000000
      25%
                 1.000000
                                2.000000
                                              2.000000
                                                         1262.476593 40013.000000
      50%
                                                         2855.374878 40025.000000
                 2.000000
                                2.000000
                                              2.000000
      75%
                                                         6076.500732 40038.000000
                 2.000000
                                2.000000
                                              2.000000
                 3.000000
                                2.000000
                                              2.000000 77284.484375 40050.000000
      max
                    VEREP
             56136.000000
      count
      mean
                 1.504400
      std
                 0.499985
      min
                 1.000000
      25%
                 1.000000
      50%
                 2.000000
      75%
                 2.000000
                 2.000000
      max
      [8 rows x 2739 columns]
[13]: df.columns
[13]: Index(['QUESTID2', 'FILEDATE', 'CIGEVER', 'CIGOFRSM', 'CIGWILYR', 'CIGTRY',
             'CIGYFU', 'CIGMFU', 'CIGREC', 'CIG30USE',
             'POVERTY3', 'TOOLONG', 'TROUBUND', 'PDEN10', 'COUTYP4', 'MAIIN102',
             'AIIND102', 'ANALWT_C', 'VESTR', 'VEREP'],
            dtype='object', length=2741)
[14]: filtered_df=df.copy()
     Selecting only required columns from the entire dataset (This is an iterative step after performing
     EDA)
[15]: desired columns = [
          'QUESTID2', 'IRWRKSTAT', 'IREDUHIGHST2', 'INCOME', 'IRSEX', 'MJEVER',
          'PNRANYLIF', 'COUTYP4', 'MEDMJPA2', 'DSTCHR30', 'ADDPREV', 'LSD', 'PCP', L
       ↔ 'PEYOTE', 'MESC', 'PSILCY', 'ECSTMOLLY', 'KETMINESK', 'DMTAMTFXY',
       ⇔'SALVIADIV', 'HALLUCOTH'
      1
      filtered df = df[desired columns]
```

```
[16]: | #We have various hallucinogens in our dataset we can standardize the mapping_
       ⇔and remove the unwanted data
      #like "Dont know", "Refused" etc with nan values
[17]: | lsd_mapping = {
         1: 'Yes', # "Yes"
         2: 'No', # "No"
         3: 'Yes', # "Yes logically assigned"
         91: 'No', # "Never used hallucinogens"
         94: np.nan, # "Don't know" replaced with NaN
         97: np.nan # "Refused" replaced with NaN
      }
      # Apply the mapping to the 'LSD' column
      filtered_df['LSD'] = filtered_df['LSD'].map(lsd_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\1422645912.py:11:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df['LSD'] = filtered_df['LSD'].map(lsd_mapping)
[18]: #we can create similar mapping for all hallucinogens
[19]: substance_mapping = {
         1: 'Yes', # "Yes"
         2: 'No', # "No"
         91: 'No', # "Never used hallucinogens"
         94: np.nan, # "Don't know" replaced with NaN
         97: np.nan # "Refused" replaced with NaN
      }
      substance_columns = ['PCP', 'PEYOTE', 'MESC', 'PSILCY', 'ECSTMOLLY', __
      ⇔'KETMINESK', 'DMTAMTFXY', 'SALVIADIV', 'HALLUCOTH']
      for col in substance_columns:
         filtered_df[col] = filtered_df[col].map(substance_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\497884309.py:12:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
```

```
filtered_df[col] = filtered_df[col].map(substance_mapping)
```

```
[20]: # Since we have a lot of hallucinogens we can generalize the use of them by creating a new column which will have yes if any one of the #hallucinogens in dataset is used
```

```
[21]: hallucinogen_columns = ['LSD', 'PCP', 'PEYOTE', 'MESC', 'PSILCY', 'ECSTMOLLY', \subseteq 'KETMINESK', 'DMTAMTFXY', 'SALVIADIV', 'HALLUCOTH']

filtered_df['hallucinogens'] = filtered_df[hallucinogen_columns].apply(
    lambda row: 'Yes' if 'Yes' in row.values else ('No' if all(val == 'No' for \subseteq val in row.values) else np.nan), axis=1
)
```

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\3382138147.py:3:
SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy filtered_df['hallucinogens'] = filtered_df[hallucinogen_columns].apply(

```
[22]: # Now only considering hallucination column created in filtered_df
desired_columns = [
    'QUESTID2', 'IRWRKSTAT', 'IREDUHIGHST2', 'INCOME', 'IRSEX', 'MJEVER',
    'PNRANYLIF', 'COUTYP4', 'MEDMJPA2', 'DSTCHR30', 'ADDPREV', 'hallucinogens']
filtered_df = filtered_df[desired_columns]
```

0.0.15 Data Cleaning

[23]: filtered_df.head()

[23]:		QUESTID2	IRWRKSTAT	IREDUHIGHST2	INCOME	IRSEX	MJEVER	PNRANYLIF	\
	0	43295143	1	11	4	1	1	1	
	1	65095143	1	11	4	2	2	1	
	2	49405143	1	11	4	1	1	1	
	3	51015143	4	6	1	2	2	1	
	4	31825143	4	7	4	1	2	2	

	COUTYP4	MEDMJPA2	DSTCHR30	ADDPREV	hallucinogens
0	2	2	3	1	No
1	2	1	5	1	No
2	2	1	5	2	No
3	2	1	4	2	No
4	2	1	4	1	No

0.0.16 Handling Special codes like Bad Data, Legitimate Skip etc in depression column which does not add value to the analysis

Replace special codes which does not add value with NaN values 85: 'Bad Data', 97: 'Refused', 98: 'Blank', 99: 'Legitimate Skip'

```
[24]: filtered_df['ADDPREV'].replace([85, 97, 98, 99], np.nan, inplace=True)
```

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\2321724339.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

filtered_df['ADDPREV'].replace([85, 97, 98, 99], np.nan, inplace=True)

```
[25]: filtered_df['ADDPREV'].unique()
```

```
[25]: array([ 1., 2., nan, 94.])
```

In the IRWRKSTAT (EMPLOYMENT STATUS) column - 99 indicates 12-14 year olds, this data might not be useful for analysis hence we can remove this data as these people might also be having some income, since we cannot say the exact income level it is better to drop these values

```
[26]: filtered_df['IRWRKSTAT'].replace([99], np.nan, inplace=True)
```

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\753615126.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This implace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
filtered_df['IRWRKSTAT'].replace([99], np.nan, inplace=True)
```

Since we have more than 11 levels of education in the dataset, we can categorize the education levels into 3 different levels which make more sense i.e Primary, Secondary and Higher education based on the grade studying

```
[27]: # Define a function to categorize the education levels
      def categorize_education(value):
          if value < 7:</pre>
              return 'primary education'
          elif value in [8, 9]:
              return 'intermediate education'
          else:
              return 'higher education'
      filtered_df['IREDUHIGHST2'] = filtered_df['IREDUHIGHST2'].
       →apply(categorize education)
[28]: filtered_df.head()
[28]:
         QUESTID2 IRWRKSTAT
                                   IREDUHIGHST2 INCOME IRSEX
                                                                MJEVER PNRANYLIF
      0 43295143
                         1.0
                               higher education
                                                             1
                                                                     1
                                                                                 1
      1 65095143
                               higher education
                                                             2
                                                                     2
                         1.0
                                                                                 1
                               higher education
      2 49405143
                         1.0
                                                      4
                                                             1
                                                                     1
                                                                                 1
      3 51015143
                         4.0 primary education
                                                      1
                                                             2
                                                                     2
                                                                                 1
      4 31825143
                         4.0
                               higher education
                                                             1
                                                                                2
         COUTYP4 MEDMJPA2 DSTCHR30 ADDPREV hallucinogens
```

MJEVER contains 94 and 97 which are dont know and refused to answer, these rows are not useful in our analysis hence we can exclude this data

```
[29]: filtered_df['MJEVER'].replace([94, 97], np.nan, inplace=True)
```

1.0

1.0

2.0

2.0

1.0

No

No

No

No

No

2

1

1

3

5

5

4

4

0

1

2

3

4

2

2

2

2

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\3512628858.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
filtered_df['MJEVER'].replace([94, 97], np.nan, inplace=True)
```

[30]: #Pain releiver use in lifetime 5 indicates logically assigned yes which means⊔

→ this value is logically made as yes, so we can replace this

#with 1 in the final data

[31]: filtered_df['PNRANYLIF'].replace(5, 1, inplace=True)

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\422161004.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

filtered_df['PNRANYLIF'].replace(5, 1, inplace=True)

[32]: #replcaing values which are not useful with NA filtered_df['PNRANYLIF'].replace([94, 97, 98], np.nan, inplace=True)

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\3256104701.py:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

filtered_df['PNRANYLIF'].replace([94, 97, 98], np.nan, inplace=True)

In the column How often do you feel sad, there are many columns which cannot be used for analysis we can replace them with NaN

[33]: #replcaing values which are not useful with NA filtered_df['DSTCHR30'].replace([85, 94, 97, 98, 99], np.nan, inplace=True)

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\1097070963.py:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

filtered_df['DSTCHR30'].replace([85, 94, 97, 98, 99], np.nan, inplace=True)

```
[34]: filtered_df.head()
[34]:
         QUESTID2
                                     IREDUHIGHST2
                                                  INCOME
                                                            IRSEX
                    IRWRKSTAT
                                                                   MJEVER
                                                                           PNRANYLIF
      0 43295143
                          1.0
                                higher education
                                                         4
                                                                1
                                                                       1.0
                                                                                  1.0
      1 65095143
                          1.0
                                higher education
                                                         4
                                                                2
                                                                       2.0
                                                                                  1.0
                          1.0
                                                         4
                                                                       1.0
      2 49405143
                                higher education
                                                                1
                                                                                  1.0
      3 51015143
                          4.0
                               primary education
                                                         1
                                                                2
                                                                       2.0
                                                                                  1.0
                                                                       2.0
      4 31825143
                          4.0
                                higher education
                                                         4
                                                                1
                                                                                  2.0
         COUTYP4
                  MEDMJPA2 DSTCHR30
                                       ADDPREV hallucinogens
               2
                          2
      0
                                   3.0
                                            1.0
                                                            Nο
               2
                                   5.0
      1
                          1
                                            1.0
                                                            Nο
      2
               2
                          1
                                   5.0
                                            2.0
                                                            Nο
      3
               2
                          1
                                   4.0
                                            2.0
                                                            No
               2
                          1
                                   4.0
                                            1.0
                                                            No
[35]: #Removal of duplicates from the data frame
      filtered_df.drop_duplicates(inplace=True)
[36]:
      #Set display options to have 2 decimals to have proper scale for future_
       \hookrightarrow operations
      pd.options.display.float_format = '{:.2f}'.format
[37]: filtered_df.describe()
[37]:
               QUESTID2 IRWRKSTAT
                                       INCOME
                                                 IRSEX
                                                          MJEVER PNRANYLIF COUTYP4
               56136.00
                           49581.00 56136.00 56136.00 56097.00
                                                                    55670.00 56136.00
      count
      mean 54346070.01
                               2.22
                                         2.72
                                                   1.52
                                                            1.56
                                                                        1.49
                                                                                 1.75
      std
            25631667.47
                               1.30
                                         1.14
                                                   0.50
                                                            0.50
                                                                        0.50
                                                                                 0.76
                               1.00
                                         1.00
                                                   1.00
                                                            1.00
                                                                        1.00
                                                                                 1.00
      min
            10009454.00
      25%
                                         2.00
            31982452.50
                               1.00
                                                   1.00
                                                            1.00
                                                                        1.00
                                                                                 1.00
                               2.00
                                         3.00
                                                   2.00
                                                            2.00
                                                                                 2.00
      50%
            54039390.00
                                                                        1.00
      75%
            76251052.50
                               4.00
                                         4.00
                                                   2.00
                                                            2.00
                                                                        2.00
                                                                                 2.00
      max
            99996688.00
                               4.00
                                         4.00
                                                   2.00
                                                            2.00
                                                                        2.00
                                                                                 3.00
             MEDMJPA2 DSTCHR30 ADDPREV
      count
             56136.00 42411.00 42508.00
                  1.30
                            4.37
                                      1.89
      mean
                  0.46
                            0.98
                                      4.59
      std
      min
                  1.00
                            1.00
                                      1.00
```

```
25%
                 1.00
                           4.00
                                     1.00
      50%
                 1.00
                           5.00
                                     2.00
      75%
                 2.00
                           5.00
                                     2.00
                 2.00
                           5.00
                                   94.00
      max
[38]:
      education_mapping = {
          'higher education': 3,
          'primary education': 1,
          'intermediate education': 2
      }
      filtered_df['IREDUHIGHST2'] = filtered_df['IREDUHIGHST2'].
       →replace(education_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\4283032743.py:6:
     FutureWarning: Downcasting behavior in `replace` is deprecated and will be
     removed in a future version. To retain the old behavior, explicitly call
     `result.infer_objects(copy=False)`. To opt-in to the future behavior, set
     `pd.set_option('future.no_silent_downcasting', True)`
       filtered_df['IREDUHIGHST2'] =
     filtered_df['IREDUHIGHST2'].replace(education_mapping)
[39]: filtered_df.head()
[39]:
         QUESTID2
                   IRWRKSTAT
                              IREDUHIGHST2
                                             INCOME
                                                     IRSEX
                                                            MJEVER
                                                                    PNRANYLIF \
      0 43295143
                        1.00
                                                         1
                                                              1.00
                                                                          1.00
      1 65095143
                        1.00
                                          3
                                                  4
                                                         2
                                                              2.00
                                                                          1.00
      2 49405143
                        1.00
                                          3
                                                              1.00
                                                  4
                                                         1
                                                                          1.00
      3 51015143
                        4.00
                                          1
                                                  1
                                                         2
                                                              2.00
                                                                          1.00
      4 31825143
                        4.00
                                          3
                                                  4
                                                         1
                                                              2.00
                                                                          2.00
         COUTYP4 MEDMJPA2 DSTCHR30 ADDPREV hallucinogens
                         2
      0
               2
                                3.00
                                          1.00
                                                          No
      1
               2
                         1
                                5.00
                                          1.00
                                                          No
      2
               2
                         1
                                5.00
                                          2.00
                                                          No
               2
      3
                         1
                                4.00
                                          2.00
                                                          No
               2
                         1
                                4.00
                                          1.00
                                                          No
[40]: filtered_df.shape
[40]: (56136, 12)
[41]: #Since we have already replaced unnecessary values with NaN we can drop these
       ⇔values as they are no longer useful
[42]: null_values = filtered_df.isnull().sum()
      print(null_values)
```

```
QUESTID2
     IRWRKSTAT
                       6555
     IREDUHIGHST2
                           0
     INCOME
                          0
     IRSEX
                          0
     MJEVER
                         39
     PNRANYLIF
                         466
     COUTYP4
                          0
     MEDMJPA2
                          0
                      13725
     DSTCHR30
     ADDPREV
                       13628
                        245
     hallucinogens
     dtype: int64
[43]: filtered_df_cleaned = filtered_df.dropna(subset=['IRWRKSTAT', 'DSTCHR30', |
       ⇔'ADDPREV', 'PNRANYLIF','MJEVER' ])
[44]: filtered_df_cleaned.shape
[44]: (42166, 12)
[45]: filtered_df_cleaned['IRWRKSTAT'].unique()
[45]: array([1., 4., 3., 2.])
[46]: irwrkstat_mapping = {
          1: 'Employed full time',
          2: 'Employed part time',
          3: 'Unemployed',
          4: 'Other (incl. not in labor force)',
          99: '12-14 year olds'
      }
      filtered_df_cleaned['IRWRKSTAT'] = filtered_df_cleaned['IRWRKSTAT'].
       →map(irwrkstat_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\2354628446.py:9:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['IRWRKSTAT'] =
     filtered_df_cleaned['IRWRKSTAT'].map(irwrkstat_mapping)
[47]: filtered_df_cleaned[['IRWRKSTAT']].head()
```

```
[47]:
                                IRWRKSTAT
      0
                       Employed full time
      1
                       Employed full time
      2
                       Employed full time
      3 Other (incl. not in labor force)
      4 Other (incl. not in labor force)
[48]: income_mapping = {
          1: 'Less than $20,000',
          2: '$20,000 - $49,999',
          3: '$50,000 - $74,999',
          4: '$75,000 or more'
      }
      filtered_df_cleaned['INCOME'] = filtered_df_cleaned['INCOME'].
       →map(income mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\4064115886.py:8:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['INCOME'] =
     filtered_df_cleaned['INCOME'].map(income_mapping)
[49]: filtered_df_cleaned[['INCOME']].head()
[49]:
                    INCOME
      0
           $75,000 or more
           $75,000 or more
      1
           $75,000 or more
      2
      3 Less than $20,000
           $75,000 or more
[50]: gender_mapping = {
          1: 'Male',
          2: 'Female'
      }
      filtered_df_cleaned['IRSEX'] = filtered_df_cleaned['IRSEX'].map(gender_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\4015784729.py:6:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row indexer,col indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
```

```
filtered_df_cleaned['IRSEX'].map(gender_mapping)
[51]: filtered_df_cleaned.head()
[51]:
                                                    IREDUHIGHST2 \
         QUESTID2
                                          IRWRKSTAT
      0 43295143
                                 Employed full time
                                                                 3
      1 65095143
                                 Employed full time
      2 49405143
                                 Employed full time
                                                                 3
      3 51015143 Other (incl. not in labor force)
                                                                 1
      4 31825143 Other (incl. not in labor force)
                                                                 3
                             IRSEX MJEVER PNRANYLIF COUTYP4
                                                                MEDMJPA2 DSTCHR30 \
                    INCOME
      0
           $75,000 or more
                                      1.00
                                                 1.00
                                                             2
                                                                               3.00
                              Male
                                                 1.00
                                                             2
      1
           $75,000 or more
                           Female
                                      2.00
                                                                        1
                                                                               5.00
           $75,000 or more
                              Male
                                      1.00
                                                 1.00
                                                             2
                                                                        1
                                                                               5.00
      3 Less than $20,000 Female
                                      2.00
                                                 1.00
                                                             2
                                                                        1
                                                                               4.00
           $75,000 or more
                                      2.00
                              Male
                                                 2.00
                                                             2
                                                                        1
                                                                               4.00
         ADDPREV hallucinogens
      0
            1.00
                            No
            1.00
      1
                            No
      2
            2.00
                            No
      3
            2.00
                            No
      4
            1.00
                            No
[52]: filtered_df_cleaned['IREDUHIGHST2'].unique()
[52]: array([3, 1, 2])
[53]: # Define a function to categorize the education levels
      def categorize_education(value):
          if value==1:
              return 'primary education'
          elif value==2:
              return 'High School education'
          else:
              return 'College Degree'
      filtered_df_cleaned['IREDUHIGHST2'] = filtered_df_cleaned['IREDUHIGHST2'].
       →apply(categorize_education)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\809615653.py:10:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
```

docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

filtered_df_cleaned['IRSEX'] =

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['IREDUHIGHST2'] =
     filtered_df_cleaned['IREDUHIGHST2'].apply(categorize_education)
[54]: filtered_df_cleaned['IREDUHIGHST2'].unique()
[54]: array(['College Degree', 'primary education', 'High School education'],
            dtype=object)
[55]: mj_mapping = {
          1: 'Yes',
          2: 'No'
      }
      filtered_df_cleaned['MJEVER'] = filtered_df_cleaned['MJEVER'].map(mj_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\3622425528.py:6:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['MJEVER'] = filtered_df_cleaned['MJEVER'].map(mj_mapping)
[56]: mj_mapping = {
         1.00: 'Yes',
          2.00: 'No'
      }
      filtered_df_cleaned['PNRANYLIF'] = filtered_df_cleaned['PNRANYLIF'].
       →map(mj_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\2351626243.py:7:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['PNRANYLIF'] =
     filtered_df_cleaned['PNRANYLIF'].map(mj_mapping)
[57]: metro mapping = {
          1: 'Large Metro',
          2: 'Small Metro',
          3: 'Non Metro'
```

```
}
      filtered_df_cleaned['COUTYP4'] = filtered_df_cleaned['COUTYP4'].
       →map(metro_mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\1158216921.py:8:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['COUTYP4'] =
     filtered_df_cleaned['COUTYP4'].map(metro_mapping)
[58]: mapping = {
         1: 'All of the time',
          2: 'Most of the time',
          3: 'Some of the time',
          4: 'A little of the time',
          5: 'None of the time',
          99: 'Legitimate skip'
      filtered_df_cleaned['DSTCHR30'] = filtered_df_cleaned['DSTCHR30'].map(mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\2251930963.py:9:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['DSTCHR30'] = filtered_df_cleaned['DSTCHR30'].map(mapping)
[59]: filtered_df_cleaned['ADDPREV'].unique()
[59]: array([ 1., 2., 94.])
[60]: mapping = {
          1.00: 'Yes',
          2.00: 'No',
          94.00: 'No'
      filtered_df_cleaned['ADDPREV'] = filtered_df_cleaned['ADDPREV'].map(mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\3008980501.py:6:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
```

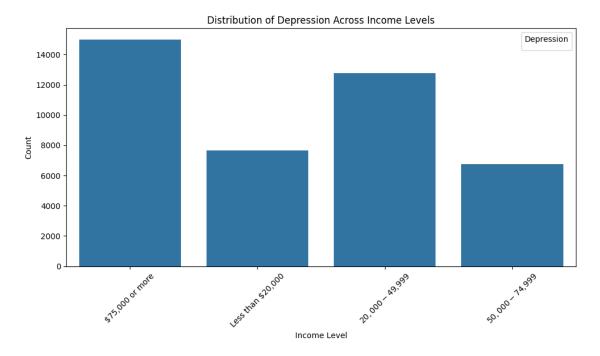
```
Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered df cleaned['ADDPREV'] = filtered df cleaned['ADDPREV'].map(mapping)
[61]: mapping = {
          1: 'Yes',
          2: 'No',
      filtered_df_cleaned['MEDMJPA2'] = filtered_df_cleaned['MEDMJPA2'].map(mapping)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\88440773.py:5:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
       filtered_df_cleaned['MEDMJPA2'] = filtered_df_cleaned['MEDMJPA2'].map(mapping)
 []:
     Renaming columns for better readability
[62]: column_mapping = {
          'IRWRKSTAT': 'Employment',
          'IREDUHIGHST2': 'education',
          'INCOME': 'income',
          'IRSEX': 'sexual orientation',
          'MJEVER': 'Ever used marijuana',
          'PNRANYLIF': 'ANY PAIN RELIEVER USE IN LIFETIME',
          'COUTYP4': 'COUNTY METRO/NONMETRO STATUS',
          'MEDMJPA2': 'STATE MEDICAL MJ LAW PASSED AT TIME OF INTERVIEW',
          'DSTCHR30': 'HOW OFTEN FELT SAD NOTHING COULD CHEER YOU UP',
          'ADDPREV': 'SEVERAL DAYS OR LNGR WHEN FELT SAD/EMPTY/DPRSD'
      }
      filtered_df_cleaned.rename(columns=column_mapping, inplace=True)
     C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\3523334497.py:14:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       filtered df_cleaned.rename(columns=column_mapping, inplace=True)
[63]: filtered_df_cleaned.to_excel('filtered_df_cleaned.xlsx')
```

```
[64]: filtered_df_cleaned.head()
[64]:
         QUESTID2
                                          Employment
                                                               education \
      0 43295143
                                  Employed full time
                                                          College Degree
      1 65095143
                                  Employed full time
                                                          College Degree
      2 49405143
                                  Employed full time
                                                          College Degree
      3 51015143 Other (incl. not in labor force)
                                                      primary education
      4 31825143 Other (incl. not in labor force)
                                                          College Degree
                    income sexual orientation Ever used marijuana
      0
           $75,000 or more
                                          Male
                                                                Yes
      1
           $75,000 or more
                                        Female
                                                                 No
      2
           $75,000 or more
                                          Male
                                                                Yes
      3
        Less than $20,000
                                        Female
                                                                 No
           $75,000 or more
                                          Male
                                                                 No
        ANY PAIN RELIEVER USE IN LIFETIME COUNTY METRO/NONMETRO STATUS
                                       Yes
                                                             Small Metro
                                       Yes
                                                             Small Metro
      1
                                                             Small Metro
      2
                                       Yes
      3
                                       Yes
                                                             Small Metro
      4
                                                             Small Metro
                                        No
        STATE MEDICAL MJ LAW PASSED AT TIME OF INTERVIEW
      0
                                                        No
                                                       Yes
      1
      2
                                                       Yes
      3
                                                       Yes
      4
                                                       Yes
        HOW OFTEN FELT SAD NOTHING COULD CHEER YOU UP
      0
                                      Some of the time
      1
                                      None of the time
      2
                                      None of the time
                                  A little of the time
      3
      4
                                  A little of the time
        SEVERAL DAYS OR LNGR WHEN FELT SAD/EMPTY/DPRSD hallucinogens
      0
                                                     Yes
      1
                                                     Yes
                                                                    No
      2
                                                     No
                                                                    No
      3
                                                     No
                                                                    No
                                                     Yes
                                                                    No
[65]: plt.figure(figsize=(10, 6))
      sns.countplot(data=filtered_df_cleaned, x='income')
      plt.title('Distribution of Depression Across Income Levels')
```

```
plt.xlabel('Income Level')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.legend(title='Depression')
plt.tight_layout()
plt.show()
```

C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\1276217604.py:7: UserWarning: No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.

plt.legend(title='Depression')



```
[]:

[66]: # Question 1: How does socioeconomic status (income, education, employment status) influence mental health?

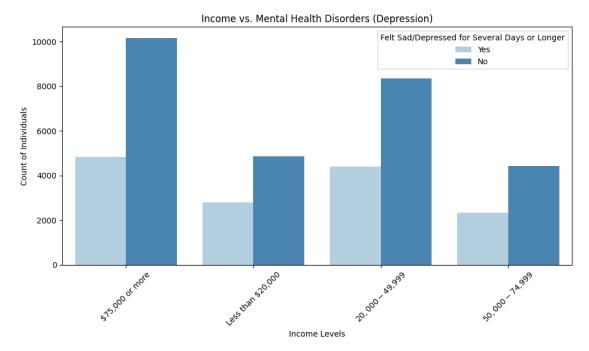
# Bar plot to show the distribution of depression (Sadness) across different sincome levels.

# Create a plot for the influence of income on experiencing mental health sistered in sincome (depression)

plt.figure(figsize=(10, 6))

sns.countplot(data=filtered_df_cleaned, x='income', hue='SEVERAL DAYS OR LNGR when Felt SAD/EMPTY/DPRSD', palette="Blues")
```

```
plt.title('Income vs. Mental Health Disorders (Depression)')
plt.xlabel('Income Levels')
plt.ylabel('Count of Individuals')
plt.xticks(rotation=45)
plt.legend(title='Felt Sad/Depressed for Several Days or Longer')
plt.tight_layout()
plt.show()
```



0.0.17 Observations:

- 1 From the above graphs we might conclude that people whose income is less than \$20,000 (which is the least among the others) are more likely to
- 2 feel depressed.

```
[67]: # Question 2: The role of marijuana use in mental health

# We will create a bar plot to see the relation between marijuana use, state

→ marijuana laws, and depression

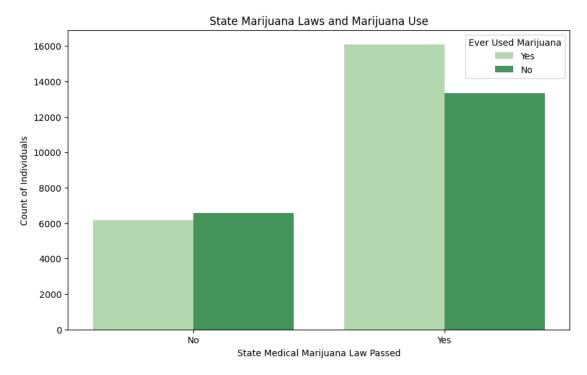
plt.figure(figsize=(10, 6))

sns.countplot(data=filtered_df_cleaned, x='STATE MEDICAL MJ LAW PASSED AT TIME

→ OF INTERVIEW', hue='Ever used marijuana', palette="Greens")

plt.title('State Marijuana Laws and Marijuana Use')
```

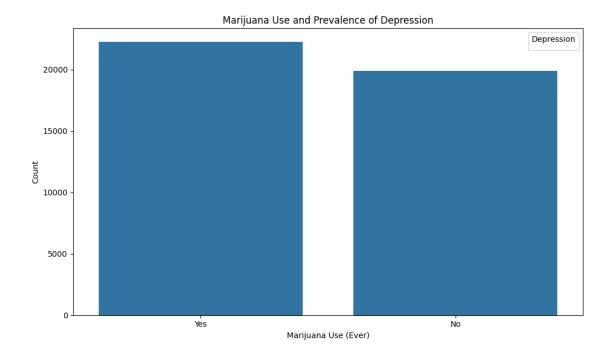
```
plt.xlabel('State Medical Marijuana Law Passed')
plt.ylabel('Count of Individuals')
plt.legend(title='Ever Used Marijuana')
#plt.tight_layout()
plt.show()
```



```
[68]: # Relationship between marijuana use and depression
   plt.figure(figsize=(10, 6))
   sns.countplot(data=filtered_df_cleaned, x='Ever used marijuana')
   plt.title('Marijuana Use and Prevalence of Depression')
   plt.xlabel('Marijuana Use (Ever)')
   plt.ylabel('Count')
   plt.legend(title='Depression')
   plt.tight_layout()
   plt.show()
```

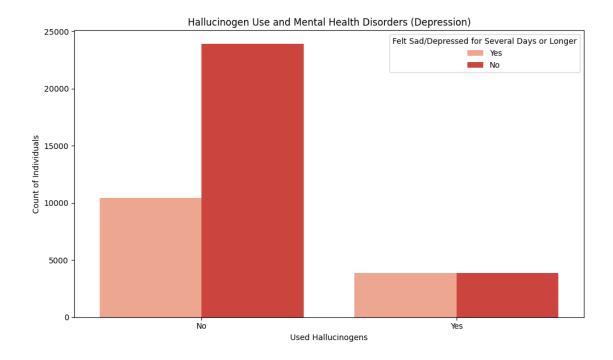
C:\Users\Rama Rao\AppData\Local\Temp\ipykernel_15216\1028580500.py:7:
UserWarning: No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.

plt.legend(title='Depression')



2.0.1 Observation:

We have observed that the locations where marijuana is legalized have higher usage of marijuana when compared to areas with stricter laws, and these regions with increased marijuana use also exhibit a higher prevalence of depression.



3.0.1 Observation: The above graph clearly shows that the percentage of people experiencing depression is higher among those who use hallucinogens compared to those who do not use them.