

CMS_Fraud_EDA

December 17, 2025

0.1 CMS Exclusions Dataset Exploratory Data Analysis

```
[170]: # Import necessary libraries such as numpy and pandas
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import os
```

```
[171]: # This function will provide basic insights into any data
def quick_summary(df):
    print("=== Missing Values ===")
    print(df.isnull().sum())
    print("\n=== Basic Info ===")
    print(df.info())
    print("\n=== Sample Rows ===")
    print(df.head())
```

0.1.1 OIG Exclusions Data

```
[172]: # Load the OIG exclusions file into a pandas dataframe
exclusions_file = 'cms_data/leie.csv'
exclusions = pd.read_csv(exclusions_file)
exclusions = exclusions[exclusions.NPI > 0].set_index('NPI')
```

```
/tmp/ipykernel_98/1029917934.py:3: DtypeWarning: Columns (3) have mixed types.
Specify dtype option on import or set low_memory=False.
exclusions = pd.read_csv(exclusions_file)
```

```
[173]: quick_summary(exclusions)
```

```
=== Missing Values ===
LASTNAME      524
FIRSTNAME     525
MIDNAME       2475
BUSNAME       7773
GENERAL        0
SPECIALTY     117
UPIN          6897
```

```

DOB                524
ADDRESS            0
CITY               0
STATE             0
ZIP               0
EXCLTYPE          0
EXCLDATE          0
REINDATE          0
WAIVERDATE        0
WVRSTATE          8293
dtype: int64

```

=== Basic Info ===

```

<class 'pandas.core.frame.DataFrame'>
Index: 8297 entries, 1972902351 to 1831242650
Data columns (total 17 columns):
#   Column          Non-Null Count  Dtype
---  -
0   LASTNAME        7773 non-null   object
1   FIRSTNAME       7772 non-null   object
2   MIDNAME         5822 non-null   object
3   BUSNAME         524 non-null    object
4   GENERAL         8297 non-null   object
5   SPECIALTY       8180 non-null   object
6   UPIN            1400 non-null   object
7   DOB             7773 non-null   float64
8   ADDRESS         8297 non-null   object
9   CITY            8297 non-null   object
10  STATE           8297 non-null   object
11  ZIP             8297 non-null   int64
12  EXCLTYPE        8297 non-null   object
13  EXCLDATE        8297 non-null   int64
14  REINDATE        8297 non-null   int64
15  WAIVERDATE      8297 non-null   int64
16  WVRSTATE        4 non-null      object
dtypes: float64(1), int64(4), object(12)
memory usage: 1.1+ MB
None

```

=== Sample Rows ===

	LASTNAME	FIRSTNAME	MIDNAME	BUSNAME \
NPI				
1972902351	NaN	NaN	NaN	101 FIRST CARE PHARMACY INC
1922348218	NaN	NaN	NaN	184TH STREET PHARMACY CORP
1942476080	NaN	NaN	NaN	A & Y MEDICAL SUPPLY, INC
1275600959	NaN	NaN	NaN	A CARING ALTERNATIVE, INC
1891731758	NaN	NaN	NaN	A FAIR DEAL PHARMACY, INC

	GENERAL	SPECIALTY	UPIN	DOB	\
NPI					
1972902351	OTHER BUSINESS	PHARMACY	NaN	NaN	
1922348218	OTHER BUSINESS	PHARMACY	NaN	NaN	
1942476080	DME COMPANY	DME - GENERAL	NaN	NaN	
1275600959	OTHER BUSINESS	HOME HEALTH AGENCY	NaN	NaN	
1891731758	OTHER BUSINESS	PHARMACY	NaN	NaN	

	ADDRESS	CITY	STATE	ZIP	\
NPI					
1972902351	C/O 609 W 191ST STREET, APT D	NEW YORK	NY	10040	
1922348218	69 E 184TH ST	BRONX	NY	10468	
1942476080	6310 108TH STREET, APT 6J	FOREST HILLS	NY	11375	
1275600959	1229 HURON RD E, FLR 6TH	CLEVELAND	OH	44115	
1891731758	C/O P O BOX 329014, #69709-05	BROOKLYN	NY	11232	

	EXCLTYPE	EXCLDATE	REINDATE	WAIVERDATE	WVRSTATE
NPI					
1972902351	1128b8	20220320	0	0	NaN
1922348218	1128a1	20180419	0	0	NaN
1942476080	1128b8	20170518	0	0	NaN
1275600959	1128a1	20130320	0	0	NaN
1891731758	1128b8	20170518	0	0	NaN

```
[174]: # Drops unimportant columns or columns with mostly missing values
exclusions = exclusions.drop(
    columns=['UPIN', 'MIDNAME', 'WVRSTATE', 'BUSNAME', 'REINDATE',
    ↪ 'WAIVERDATE'],
    errors='ignore'
)

exclusions.head()
```

```
[174]:
```

	LASTNAME	FIRSTNAME	GENERAL	SPECIALTY	DOB	\
NPI						
1972902351	NaN	NaN	OTHER BUSINESS	PHARMACY	NaN	
1922348218	NaN	NaN	OTHER BUSINESS	PHARMACY	NaN	
1942476080	NaN	NaN	DME COMPANY	DME - GENERAL	NaN	
1275600959	NaN	NaN	OTHER BUSINESS	HOME HEALTH AGENCY	NaN	
1891731758	NaN	NaN	OTHER BUSINESS	PHARMACY	NaN	

	ADDRESS	CITY	STATE	ZIP	\
NPI					
1972902351	C/O 609 W 191ST STREET, APT D	NEW YORK	NY	10040	
1922348218	69 E 184TH ST	BRONX	NY	10468	
1942476080	6310 108TH STREET, APT 6J	FOREST HILLS	NY	11375	
1275600959	1229 HURON RD E, FLR 6TH	CLEVELAND	OH	44115	

1891731758 C/O P O BOX 329014, #69709-05 BROOKLYN NY 11232

	EXCLTYPE	EXCLDATE
NPI		
1972902351	1128b8	20220320
1922348218	1128a1	20180419
1942476080	1128b8	20170518
1275600959	1128a1	20130320
1891731758	1128b8	20170518

```
[175]: exclusions = exclusions.dropna(subset=['SPECIALTY'])
exclusions = exclusions.dropna(subset=['DOB'])
print(exclusions.shape)
```

(7723, 11)

```
[176]: date_cols = ['DOB', 'EXCLDATE']
for col in date_cols:
    exclusions[col] = exclusions[col].astype(str).str.replace(r'\.0$', '',
    regex=True)
    exclusions[col] = exclusions[col].replace(['nan', 'NaN', '', '00000000'],
    np.nan)
    exclusions[col] = pd.to_datetime(exclusions[col], format='%Y%m%d',
    errors='coerce')

exclusions.head()
```

```
[176]:
```

	LASTNAME	FIRSTNAME	GENERAL	SPECIALTY	\
NPI					
1760461826	ABAD-SANTOS	CRISELDA	PHYSICIAN (MD, DO)	PSYCHIATRY	
1477537496	ABADI	JAMSHEED	PHYSICIAN (MD, DO)	INTERNAL MEDICINE	
1124292966	ABARIENTOS	CRISPIN	PHYSICIAN (MD, DO)	RHEUMATOLOGY	
1376108431	ABBAS	SHAFI	BUS OWNER/EXEC	DME - PROSTHETICS	
1194807255	ABBASSI	JADAN	PHYSICIAN (MD, DO)	GENERAL PRACTICE	

	DOB	ADDRESS	CITY	STATE	ZIP	\
NPI						
1760461826	1963-12-20	8506 N ADIR DR	WEST HILLS	CA	91304	
1477537496	1939-01-10	89 WEEKS ROAD	E WILLISTON PARK	NY	11596	
1124292966	1974-09-19	P O BOX 879, #26401-014	AYER	MA	1432	
1376108431	1967-06-06	P O BOX 26020	BEAUMONT	TX	26020	
1194807255	1944-09-19	115 NELLIS DRIVE	WAYNE	NJ	7470	

	EXCLTYPE	EXCLDATE
NPI		
1760461826	1128b4	2025-01-20
1477537496	1128b4	2014-05-20

```

1124292966    1128a1 2020-06-18
1376108431    1128a1 2025-10-20
1194807255    1128b4 2018-06-20

```

```

[177]: exclusions['age_at_exclusion'] = (
        (exclusions['EXCLDATE'] - exclusions['DOB']).dt.days / 365.25
    )
    exclusions['year_excl'] = exclusions['EXCLDATE'].dt.year

```

```

[178]: excltype_map = {
        '1128a1': 'Conviction - Medicare Fraud',
        '1128a2': 'Patient Abuse/Neglect',
        '1128a3': 'Felony - Drugs',
        '1128a4': 'Felony - Healthcare Fraud',
        '1128b1': 'Misdemeanor - Fraud',
        '1128b2': 'Default on Student Loan',
        '1128b3': 'License Revocation',
        '1128b4': 'Unlawful Claims',
        '1128b5': 'Kickbacks/Bribery',
        '1128b6': 'False Claims',
        '1128b7': 'Obstruction of Audit',
        '1128b8': 'Controlled Substances Violation',
        '1128b9': 'Insurance Fraud',
        '1128b10': 'Unlawful Billing',
        '1128b11': 'Quality of Care Violation',
        '1128b12': 'Civil Monetary Penalty',
        '1128b13': 'False Statement',
        '1128b14': 'Suspension/Exclusion',
        '1128b15': 'License Suspension',
        '1128b16': 'Federal Program Violation',
    }

    exclusions['EXCLTYPE'] = exclusions['EXCLTYPE'].replace(excltype_map)

    exclusions['EXCLTYPE'] = exclusions['EXCLTYPE'].fillna('Other')

    print(exclusions['EXCLTYPE'].value_counts())

```

```

EXCLTYPE
Conviction - Medicare Fraud    2921
Unlawful Claims                2388
Felony - Healthcare Fraud      982
Felony - Drugs                 642
Patient Abuse/Neglect          370
Suspension/Exclusion            197
Obstruction of Audit            84
Kickbacks/Bribery              51

```

```

Misdemeanor - Fraud          41
License Revocation           20
False Claims                  7
Default on Student Loan       7
1128Aa                        4
BRCH SA                       4
BRCH CIA                      3
Federal Program Violation     1
1156                          1
Name: count, dtype: int64

```

```
[179]: quick_summary(exclusions)
```

```
=== Missing Values ===
```

```

LASTNAME      0
FIRSTNAME     1
GENERAL       0
SPECIALTY     0
DOB           0
ADDRESS       0
CITY          0
STATE         0
ZIP           0
EXCLTYPE      0
EXCLDATE      0
age_at_exclusion 0
year_excl     0
dtype: int64

```

```
=== Basic Info ===
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 7723 entries, 1760461826 to 1831242650
```

```
Data columns (total 13 columns):
```

#	Column	Non-Null Count	Dtype
0	LASTNAME	7723 non-null	object
1	FIRSTNAME	7722 non-null	object
2	GENERAL	7723 non-null	object
3	SPECIALTY	7723 non-null	object
4	DOB	7723 non-null	datetime64[ns]
5	ADDRESS	7723 non-null	object
6	CITY	7723 non-null	object
7	STATE	7723 non-null	object
8	ZIP	7723 non-null	int64
9	EXCLTYPE	7723 non-null	object
10	EXCLDATE	7723 non-null	datetime64[ns]
11	age_at_exclusion	7723 non-null	float64
12	year_excl	7723 non-null	int32

```
dtypes: datetime64[ns](2), float64(1), int32(1), int64(1), object(8)
memory usage: 814.5+ KB
None
```

=== Sample Rows ===

	LASTNAME	FIRSTNAME	GENERAL	SPECIALTY	\
NPI					
1760461826	ABAD-SANTOS	CRISELDA	PHYSICIAN (MD, DO)	PSYCHIATRY	
1477537496	ABADI	JAMSHEED	PHYSICIAN (MD, DO)	INTERNAL MEDICINE	
1124292966	ABARIENTOS	CRISPIN	PHYSICIAN (MD, DO)	RHEUMATOLOGY	
1376108431	ABBAS	SHAFI	BUS OWNER/EXEC	DME - PROSTHETICS	
1194807255	ABBASSI	JADAN	PHYSICIAN (MD, DO)	GENERAL PRACTICE	

	DOB	ADDRESS	CITY	STATE	ZIP	\
NPI						
1760461826	1963-12-20	8506 N ADIR DR	WEST HILLS	CA	91304	
1477537496	1939-01-10	89 WEEKS ROAD	E WILLISTON PARK	NY	11596	
1124292966	1974-09-19	P O BOX 879, #26401-014	AYER	MA	1432	
1376108431	1967-06-06	P O BOX 26020	BEAUMONT	TX	26020	
1194807255	1944-09-19	115 NELLIS DRIVE	WAYNE	NJ	7470	

	EXCLTYPE	EXCLDATE	age_at_exclusion	\
NPI				
1760461826	Unlawful Claims	2025-01-20	61.086927	
1477537496	Unlawful Claims	2014-05-20	75.356605	
1124292966	Conviction - Medicare Fraud	2020-06-18	45.746749	
1376108431	Conviction - Medicare Fraud	2025-10-20	58.373717	
1194807255	Unlawful Claims	2018-06-20	73.749487	

	year_excl
NPI	
1760461826	2025
1477537496	2014
1124292966	2020
1376108431	2025
1194807255	2018

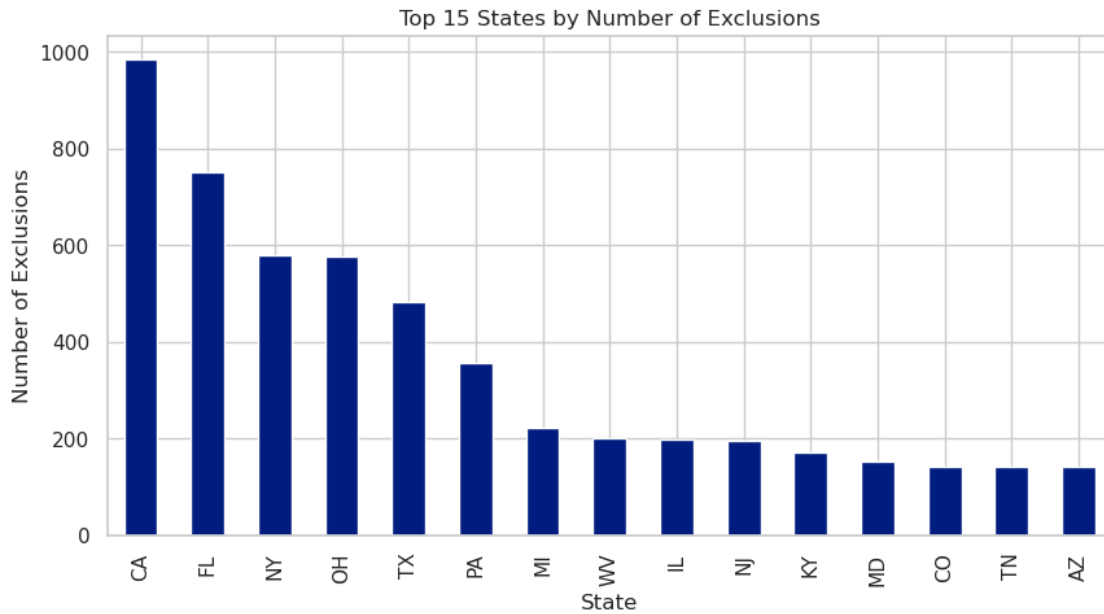
```
[180]: exclusions['SPECIALTY'].value_counts().head(15)
```

```
[180]: SPECIALTY
NURSE/NURSES AIDE      826
GENERAL PRACTICE      753
FAMILY PRACTICE       591
INTERNAL MEDICINE      546
CHIROPRACTIC          422
COUNSELOR             397
DENTIST               334
```

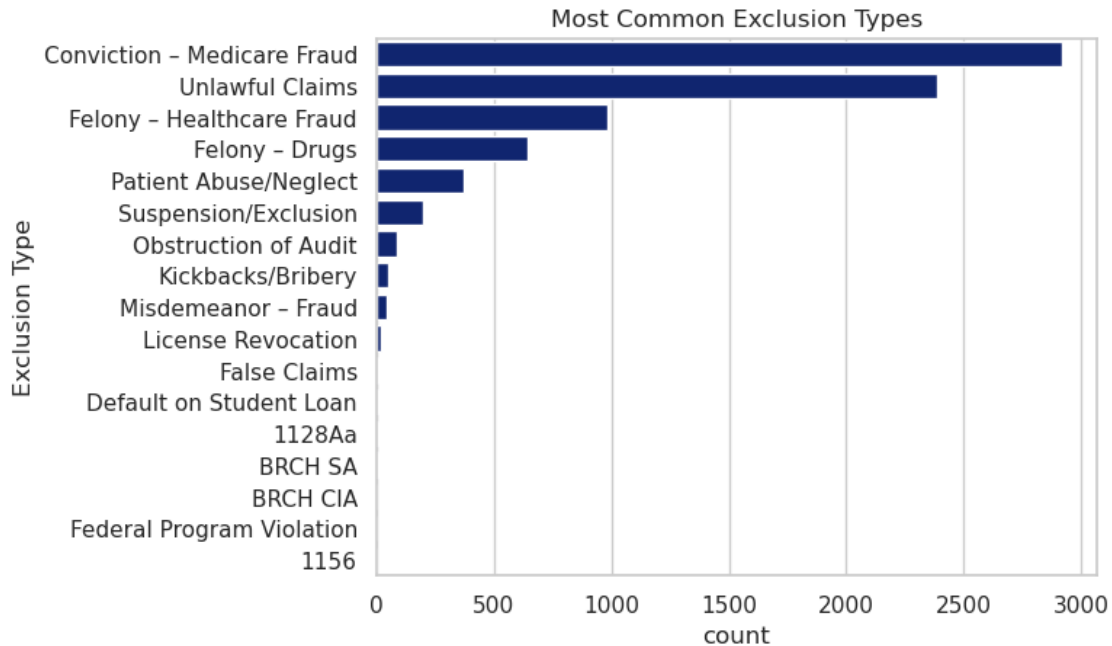
PHARMACIST	324
PSYCHIATRY	243
NURSE PRACTITIONER (188
PAIN MANAGEMENT	179
SOCIAL WORKER	172
PHYSICIAN ASSISTANT	165
THERAPIST	160
PSYCHOLOGY	151

Name: count, dtype: int64

```
[181]: plt.figure(figsize=(10,5))
exclusions['STATE'].value_counts().head(15).plot(kind='bar')
plt.title('Top 15 States by Number of Exclusions')
plt.xlabel('State')
plt.ylabel('Number of Exclusions')
plt.show()
```



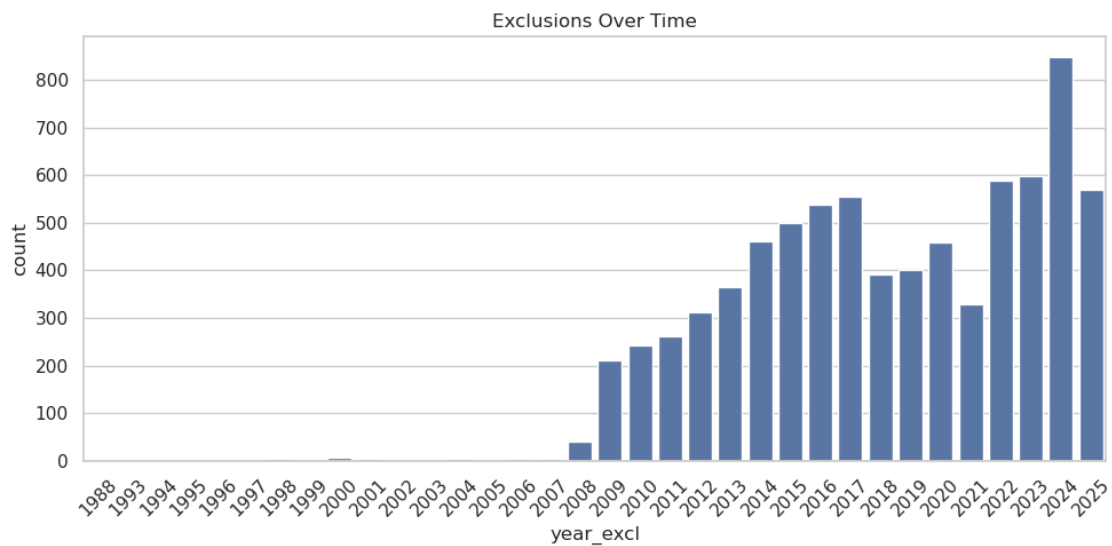
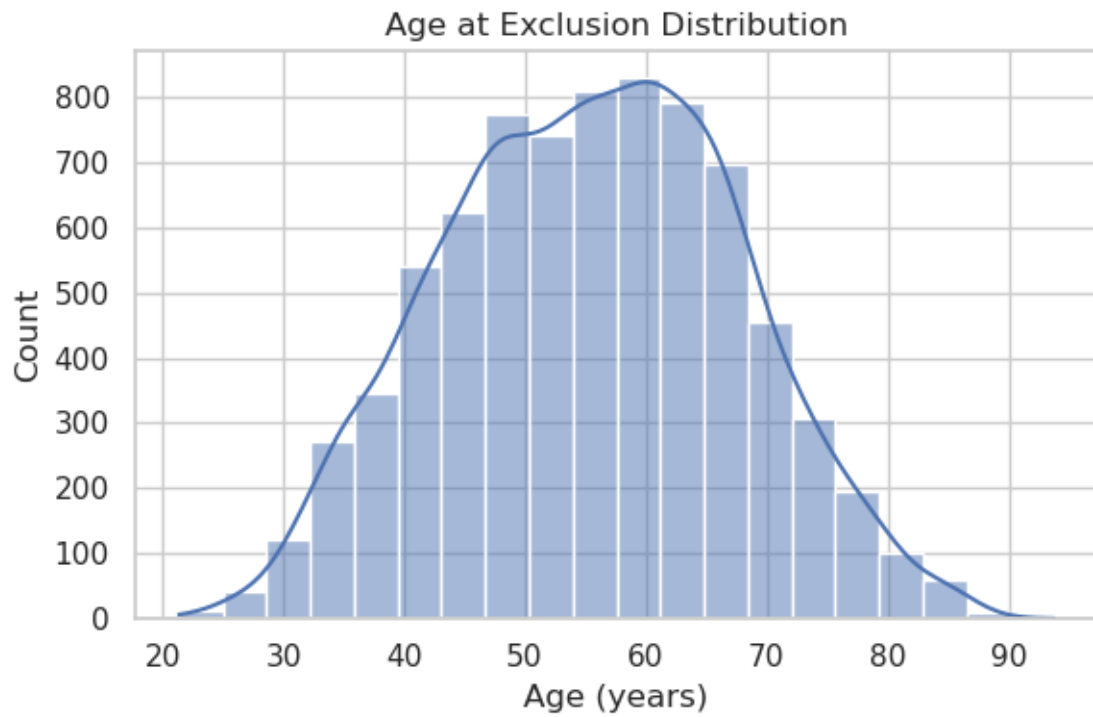
```
[182]: sns.countplot(y='EXCLTYPE', data=exclusions, order=exclusions['EXCLTYPE'].
        ↪value_counts().index)
plt.ylabel("Exclusion Type")
plt.title('Most Common Exclusion Types')
plt.show()
```

```
[183]: sns.set(style="whitegrid")

plt.figure(figsize=(6, 4))
sns.histplot(exclusions['age_at_exclusion'].dropna(), kde=True, bins=20)
plt.title("Age at Exclusion Distribution")
plt.xlabel("Age (years)")
plt.ylabel("Count")
plt.tight_layout()
plt.show()

plt.figure(figsize=(10, 5))
sns.countplot(
    data=exclusions,
    x='year_excl',
    order=sorted(exclusions['year_excl'].dropna().unique())
)
plt.title("Exclusions Over Time")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
[184]: top_states = (
    exclusions['STATE']
    .value_counts()
    .nlargest(10)
    .reset_index()
```

```

)

top_states.columns = ['STATE', 'COUNT']

plt.figure(figsize=(8, 5))
sns.barplot(data=top_states, x='STATE', y='COUNT', palette='viridis')

plt.title("Top 10 States by Medicare Exclusions", fontsize=14, weight='bold')
plt.xlabel("State")
plt.ylabel("Number of Exclusions")

for i, row in top_states.iterrows():
    plt.text(i, row['COUNT'] + 5, int(row['COUNT']), ha='center')

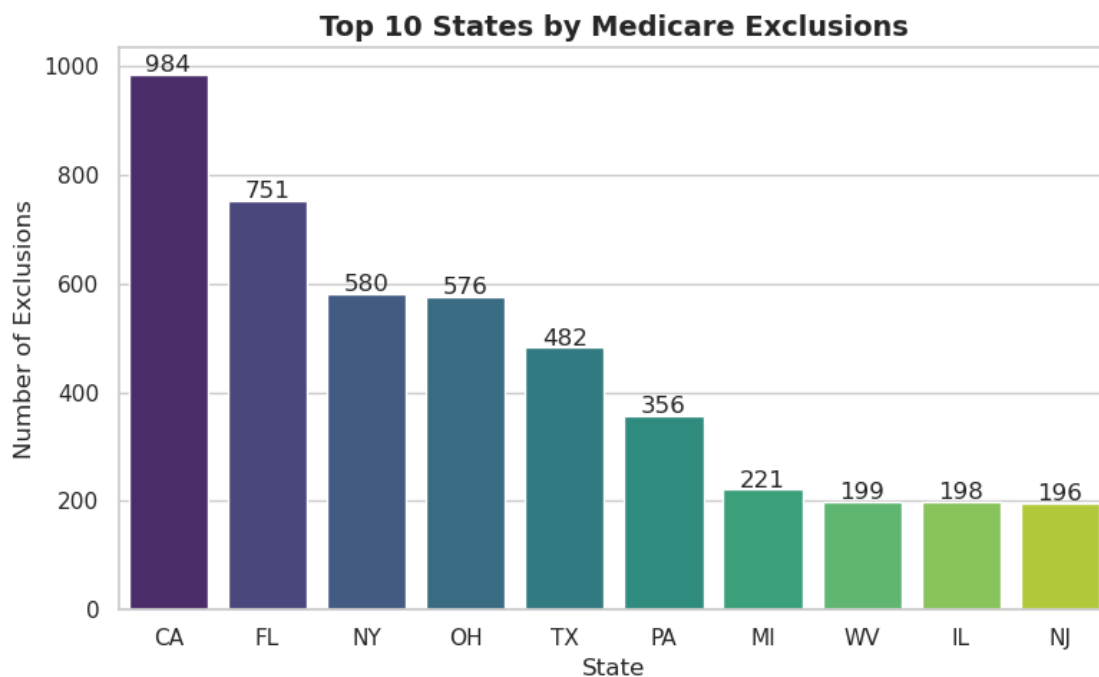
plt.tight_layout()
plt.show()

```

/tmp/ipykernel_98/3599476972.py:12: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(data=top_states, x='STATE', y='COUNT', palette='viridis')
```



0.1.2 CMS Medicare Provider Data

```
[185]: files = os.listdir('cms_data')
files
```

```
[185]: ['MUP_PHY_R25_P07_V10_D19_Prov.csv',
'MUP_PHY_R25_P07_V10_D22_Prov.csv',
'.ipynb_checkpoints',
'MUP_PHY_R25_P07_V10_D21_Prov.csv',
'leie.csv',
'y2022_prep.csv',
'y2019_prep.csv',
'MUP_PHY_R25_P05_V20_D23_Prov.csv',
'MUP_PHY_R25_P07_V10_D20_Prov.csv',
'y2023_prep.csv']
```

```
[186]: fil = files[6]
data = pd.read_csv('cms_data/'+fil)
```

```
[187]: data = data.rename(columns={'Rndrng_NPI': 'NPI'})
data = data[data['NPI'] > 0].set_index('NPI')
data= data.drop('Unnamed: 0', axis=1)
```

```
[188]: data.head()
```

```
[188]:
```

	Rndrng_Privr_Last_Org_Name	Rndrng_Privr_First_Name	Rndrng_Privr_MI	\
NPI				
1003000134	CIBULL	THOMAS	L	
1003005315	SMITH	ADAM	B	
1003009861	BANNA	MOUSTAFA	NaN	
1003010570	CHOW	LING	S	
1003017443	BLOKAR	MIRJANA	NaN	

	Rndrng_Privr_Crdntls	Rndrng_Privr_Ent_Cd	Rndrng_Privr_St1	\
NPI				
1003000134	M.D.	I	2650 RIDGE AVE	
1003005315	MD	I	4977 SKYVIEW CT	
1003009861	MD	I	5859 W. TALAVI BLVD	
1003010570	M.D.	I	900 E BROADWAY AVE	
1003017443	M.D.	I	65 BLEECKER ST	

	Rndrng_Privr_St2	Rndrng_Privr_City	Rndrng_Privr_State_Abrvtn	\
NPI				
1003000134	EVANSTON HOSPITAL	EVANSTON	IL	
1003005315	NaN	TRAVERSE CITY	MI	
1003009861	SUITE 100	GLENDALE	AZ	
1003010570	NaN	BISMARCK	ND	
1003017443	12TH FLOOR	NEW YORK	NY	

	Rndrng_Prldr_State_FIPS	Rndrng_Prldr_Zip5	Rndrng_Prldr_RUCA	\
NPI				
1003000134	17	60201.0	1.0	
1003005315	26	49684.0	4.0	
1003009861	4	85306.0	1.0	
1003010570	38	58501.0	1.0	
1003017443	36	10012.0	1.0	

	Rndrng_Prldr_RUCA_Desc	\
NPI		
1003000134	Metropolitan area core: primary flow within an...	
1003005315	Micropolitan area core: primary flow within an...	
1003009861	Metropolitan area core: primary flow within an...	
1003010570	Metropolitan area core: primary flow within an...	
1003017443	Metropolitan area core: primary flow within an...	

	Rndrng_Prldr_Cntry	Rndrng_Prldr_Type	\
NPI			
1003000134	US	Pathology	
1003005315	US	Plastic and Reconstructive Surgery	
1003009861	US	Cardiology	
1003010570	US	Internal Medicine	
1003017443	US	Psychiatry	

	Rndrng_Prldr_Mdcr_Prtcptg_Ind	Tot_HCPCS_Cds	Tot_Benes	Tot_Srvcs	\
NPI					
1003000134	Y	20	3614	7835.0	
1003005315	Y	51	71	533.0	
1003009861	Y	68	876	3581.0	
1003010570	Y	21	309	833.0	
1003017443	Y	5	64	244.0	

	Tot_Sbmted_Chrg	Tot_Mdcr_Alowd_Amt	Tot_Mdcr_Pymt_Amt	\
NPI				
1003000134	1213264.0	298905.09	228078.56	
1003005315	250569.0	86114.40	67712.67	
1003009861	788610.0	361645.81	280928.83	
1003010570	173991.0	84420.40	65891.84	
1003017443	45375.0	23956.84	18245.22	

	Tot_Mdcr_Stdzd_Amt	Drug_Sprsn_Ind	Drug_Tot_HCPCS_Cds	\
NPI				
1003000134	218268.63	NaN	0.0	
1003005315	70489.71	*	NaN	
1003009861	286768.20	NaN	1.0	
1003010570	66566.70	NaN	0.0	

1003017443	16748.76	NaN	0.0
------------	----------	-----	-----

	Drug_Tot_Benes	Drug_Tot_Srvcs	Drug_Sbmted_Chrg \
NPI			
1003000134	0.0	0.0	0.0
1003005315	NaN	NaN	NaN
1003009861	98.0	392.0	63504.0
1003010570	0.0	0.0	0.0
1003017443	0.0	0.0	0.0

	Drug_Mdcr_Alowd_Amt	Drug_Mdcr_Pymt_Amt	Drug_Mdcr_Stdzd_Amt \
NPI			
1003000134	0.00	0.00	0.00
1003005315	NaN	NaN	NaN
1003009861	21905.24	17472.35	18074.19
1003010570	0.00	0.00	0.00
1003017443	0.00	0.00	0.00

	Med_Sprsn_Ind	Med_Tot_HCPCS_Cds	Med_Tot_Benes	Med_Tot_Srvcs \
NPI				
1003000134	NaN	20.0	3614.0	7835.0
1003005315	#	NaN	NaN	NaN
1003009861	NaN	67.0	876.0	3189.0
1003010570	NaN	21.0	309.0	833.0
1003017443	NaN	5.0	64.0	244.0

	Med_Sbmted_Chrg	Med_Mdcr_Alowd_Amt	Med_Mdcr_Pymt_Amt \
NPI			
1003000134	1213264.0	298905.09	228078.56
1003005315	NaN	NaN	NaN
1003009861	725106.0	339740.57	263456.48
1003010570	173991.0	84420.40	65891.84
1003017443	45375.0	23956.84	18245.22

	Med_Mdcr_Stdzd_Amt	Bene_Avg_Age	Bene_Age_LT_65_Cnt \
NPI			
1003000134	218268.63	76	97.0
1003005315	NaN	73	NaN
1003009861	268694.01	73	91.0
1003010570	66566.70	76	39.0
1003017443	16748.76	83	NaN

	Bene_Age_65_74_Cnt	Bene_Age_75_84_Cnt	Bene_Age_GT_84_Cnt \
NPI			
1003000134	1623.0	1333.0	561.0
1003005315	29.0	20.0	NaN
1003009861	400.0	294.0	91.0

1003010570	95.0	100.0	75.0
1003017443	NaN	18.0	31.0

	Bene_Feml_Cnt	Bene_Male_Cnt	Bene_Race_Wht_Cnt \
NPI			
1003000134	1767.0	1847.0	3312.0
1003005315	44.0	27.0	NaN
1003009861	410.0	466.0	709.0
1003010570	167.0	142.0	294.0
1003017443	42.0	22.0	48.0

	Bene_Race_Black_Cnt	Bene_Race_API_Cnt	Bene_Race_Hspnc_Cnt \
NPI			
1003000134	41.0	62.0	43.0
1003005315	NaN	NaN	NaN
1003009861	35.0	20.0	80.0
1003010570	NaN	NaN	NaN
1003017443	NaN	0.0	NaN

	Bene_Race_NatInd_Cnt	Bene_Race_Othr_Cnt	Bene_Dual_Cnt \
NPI			
1003000134	0.0	156.0	195.0
1003005315	NaN	NaN	14.0
1003009861	11.0	21.0	164.0
1003010570	NaN	NaN	77.0
1003017443	0.0	NaN	NaN

	Bene_Ndual_Cnt	Bene_CC_BH_ADHD_OthCD_V1_Pct \
NPI		
1003000134	3419.0	0.0
1003005315	57.0	NaN
1003009861	712.0	NaN
1003010570	232.0	NaN
1003017443	NaN	NaN

	Bene_CC_BH_Alcohol_Drug_V1_Pct	Bene_CC_BH_Tobacco_V1_Pct \
NPI		
1003000134	2.0	3.0
1003005315	NaN	NaN
1003009861	13.0	21.0
1003010570	8.0	17.0
1003017443	NaN	NaN

	Bene_CC_BH_Alz_NonAlzdem_V2_Pct	Bene_CC_BH_Anxiety_V1_Pct \
NPI		
1003000134	4.0	13.0
1003005315	NaN	35.0

1003009861	12.0	29.0
1003010570	26.0	42.0
1003017443	63.0	45.0

	Bene_CC_BH_Bipolar_V1_Pct	Bene_CC_BH_Mood_V2_Pct \
NPI		
1003000134	2.0	16.0
1003005315	NaN	32.0
1003009861	5.0	31.0
1003010570	9.0	46.0
1003017443	NaN	75.0

	Bene_CC_BH_Depress_V1_Pct	Bene_CC_BH_PD_V1_Pct \
NPI		
1003000134	14.0	1.0
1003005315	30.0	NaN
1003009861	28.0	NaN
1003010570	45.0	NaN
1003017443	75.0	NaN

	Bene_CC_BH_PTSD_V1_Pct	Bene_CC_BH_Schizo_OthPsy_V1_Pct \
NPI		
1003000134	0.0	1.0
1003005315	NaN	NaN
1003009861	3.0	3.0
1003010570	NaN	6.0
1003017443	NaN	23.0

	Bene_CC_PH_Asthma_V2_Pct	Bene_CC_PH_Afib_V2_Pct \
NPI		
1003000134	6.0	16.0
1003005315	NaN	24.0
1003009861	15.0	38.0
1003010570	17.0	35.0
1003017443	NaN	30.0

	Bene_CC_PH_Cancer6_V2_Pct	Bene_CC_PH_CKD_V2_Pct \
NPI		
1003000134	18.0	15.0
1003005315	17.0	25.0
1003009861	19.0	42.0
1003010570	22.0	45.0
1003017443	23.0	30.0

	Bene_CC_PH_COPD_V2_Pct	Bene_CC_PH_Diabetes_V2_Pct \
NPI		
1003000134	10.0	21.0

1003005315	17.0	45.0
1003009861	34.0	40.0
1003010570	37.0	42.0
1003017443	36.0	41.0

	Bene_CC_PH_HF_NonIHD_V2_Pct	Bene_CC_PH_Hyperlipidemia_V2_Pct	\
NPI			
1003000134	11.0	72.0	
1003005315	NaN	63.0	
1003009861	37.0	75.0	
1003010570	42.0	75.0	
1003017443	44.0	75.0	

	Bene_CC_PH_Hypertension_V2_Pct	Bene_CC_PH_IschemicHeart_V2_Pct	\
NPI			
1003000134	64.0	20.0	
1003005315	75.0	23.0	
1003009861	75.0	61.0	
1003010570	75.0	49.0	
1003017443	75.0	61.0	

	Bene_CC_PH_Osteoporosis_V2_Pct	Bene_CC_PH_Parkinson_V2_Pct	\
NPI			
1003000134	15.0	3.0	
1003005315	NaN	NaN	
1003009861	13.0	2.0	
1003010570	23.0	NaN	
1003017443	20.0	NaN	

	Bene_CC_PH_Arthritis_V2_Pct	Bene_CC_PH_Stroke_TIA_V2_Pct	\
NPI			
1003000134	42.0	6.0	
1003005315	48.0	NaN	
1003009861	50.0	22.0	
1003010570	59.0	18.0	
1003017443	70.0	31.0	

	Bene_Avg_Risk_Scre	LASTNAME	FIRSTNAME	MIDNAME	BUSNAME	\
NPI						
1003000134	1.1246	NaN	NaN	NaN	NaN	
1003005315	1.6177	SMITH	ADAM	BRYANT	NaN	
1003009861	2.1074	NaN	NaN	NaN	NaN	
1003010570	2.1762	NaN	NaN	NaN	NaN	
1003017443	2.2548	NaN	NaN	NaN	NaN	

	GENERAL	SPECIALTY	UPIN	DOB	\
NPI					

1003000134	NaN	NaN	NaN	NaN
1003005315	PHYSICIAN (MD, DO)	PLASTIC SURGERY	NaN	19800519.0
1003009861	NaN	NaN	NaN	NaN
1003010570	NaN	NaN	NaN	NaN
1003017443	NaN	NaN	NaN	NaN

	ADDRESS	CITY	STATE	ZIP	EXCLTYPE \
NPI					
1003000134	NaN	NaN	NaN	NaN	UNK
1003005315	2136 FORD PKWY 5103	SAINT PAUL	MN	55116.0	1128a1
1003009861	NaN	NaN	NaN	NaN	UNK
1003010570	NaN	NaN	NaN	NaN	UNK
1003017443	NaN	NaN	NaN	NaN	UNK

	EXCLDATE	REINDATE	WAIVERDATE	WVRSTATE \
NPI				
1003000134	NaN	NaN	NaN	NaN
1003005315	20221220.0	0.0	0.0	NaN
1003009861	NaN	NaN	NaN	NaN
1003010570	NaN	NaN	NaN	NaN
1003017443	NaN	NaN	NaN	NaN

	rat_Drug_Mdcr_Alowd_Amt_Drug_Mdcr_Pymt_Amt \
NPI	
1003000134	NaN
1003005315	NaN
1003009861	1.253709
1003010570	NaN
1003017443	NaN

	rat_Tot_Mdcr_Alowd_Amt_Med_Sbmted_Chrg \
NPI	
1003000134	0.246364
1003005315	NaN
1003009861	0.498749
1003010570	0.485200
1003017443	0.527974

	rat_Drug_Tot_Benes_Tot_Benes	excluded	rand
NPI			
1003000134	0.000000	0	0.994100
1003005315	NaN	1	0.253627
1003009861	0.111872	0	0.994092
1003010570	0.000000	0	0.994872
1003017443	0.000000	0	0.991242

```
[189]: quick_summary(data)
```

=== Missing Values ===

Rndrng_Privr_Last_Org_Name	0
Rndrng_Privr_First_Name	649
Rndrng_Privr_MI	4210
Rndrng_Privr_Crdntls	1302
Rndrng_Privr_Ent_Cd	0

...

rat_Drug_Mdcr_Alowd_Amt_Drug_Mdcr_Pymt_Amt	9475
rat_Tot_Mdcr_Alowd_Amt_Med_Sbmted_Chrg	1410
rat_Drug_Tot_Benes_Tot_Benes	1410
excluded	0
rand	0

Length: 102, dtype: int64

=== Basic Info ===

<class 'pandas.core.frame.DataFrame'>

Index: 12470 entries, 1003000134 to 1992999759

Columns: 102 entries, Rndrng_Privr_Last_Org_Name to rand

dtypes: float64(70), int64(5), object(27)

memory usage: 9.8+ MB

None

=== Sample Rows ===

NPI	Rndrng_Privr_Last_Org_Name	Rndrng_Privr_First_Name	Rndrng_Privr_MI	\
1003000134	CIBULL	THOMAS	L	
1003005315	SMITH	ADAM	B	
1003009861	BANNA	MOUSTAFA	NaN	
1003010570	CHOW	LING	S	
1003017443	BLOKAR	MIRJANA	NaN	

NPI	Rndrng_Privr_Crdntls	Rndrng_Privr_Ent_Cd	Rndrng_Privr_St1	\
1003000134	M.D.	I	2650 RIDGE AVE	
1003005315	MD	I	4977 SKYVIEW CT	
1003009861	MD	I	5859 W. TALAVI BLVD	
1003010570	M.D.	I	900 E BROADWAY AVE	
1003017443	M.D.	I	65 BLEECKER ST	

NPI	Rndrng_Privr_St2	Rndrng_Privr_City	Rndrng_Privr_State_Abrvtn	\
1003000134	EVANSTON HOSPITAL	EVANSTON	IL	
1003005315	NaN	TRAVERSE CITY	MI	
1003009861	SUITE 100	GLENDALE	AZ	
1003010570	NaN	BISMARCK	ND	
1003017443	12TH FLOOR	NEW YORK	NY	

Rndrng_Privr_State_FIPS	Rndrng_Privr_Zip5	Rndrng_Privr_RUCA	\
-------------------------	-------------------	-------------------	---

NPI			
1003000134	17	60201.0	1.0
1003005315	26	49684.0	4.0
1003009861	4	85306.0	1.0
1003010570	38	58501.0	1.0
1003017443	36	10012.0	1.0

Rndrng_Privr_RUCA_Desc \

NPI	
1003000134	Metropolitan area core: primary flow within an...
1003005315	Micropolitan area core: primary flow within an...
1003009861	Metropolitan area core: primary flow within an...
1003010570	Metropolitan area core: primary flow within an...
1003017443	Metropolitan area core: primary flow within an...

Rndrng_Privr_Cntry Rndrng_Privr_Type \

NPI		
1003000134	US	Pathology
1003005315	US	Plastic and Reconstructive Surgery
1003009861	US	Cardiology
1003010570	US	Internal Medicine
1003017443	US	Psychiatry

Rndrng_Privr_Mdcr_Prtcptg_Ind Tot_HCPCS_Cds Tot_Benes Tot_Srvcs \

NPI				
1003000134	Y	20	3614	7835.0
1003005315	Y	51	71	533.0
1003009861	Y	68	876	3581.0
1003010570	Y	21	309	833.0
1003017443	Y	5	64	244.0

Tot_Sbmted_Chrg Tot_Mdcr_Alowd_Amt Tot_Mdcr_Pymt_Amt \

NPI			
1003000134	1213264.0	298905.09	228078.56
1003005315	250569.0	86114.40	67712.67
1003009861	788610.0	361645.81	280928.83
1003010570	173991.0	84420.40	65891.84
1003017443	45375.0	23956.84	18245.22

Tot_Mdcr_Stdzd_Amt Drug_Sprsn_Ind Drug_Tot_HCPCS_Cds \

NPI			
1003000134	218268.63	NaN	0.0
1003005315	70489.71	*	NaN
1003009861	286768.20	NaN	1.0
1003010570	66566.70	NaN	0.0
1003017443	16748.76	NaN	0.0

Drug_Tot_Benes Drug_Tot_Srvcs Drug_Sbmted_Chrg \

NPI			
1003000134	0.0	0.0	0.0
1003005315	NaN	NaN	NaN
1003009861	98.0	392.0	63504.0
1003010570	0.0	0.0	0.0
1003017443	0.0	0.0	0.0

	Drug_Mdcr_Alowd_Amt	Drug_Mdcr_Pymt_Amt	Drug_Mdcr_Stdzd_Amt	\
NPI				
1003000134	0.00	0.00	0.00	
1003005315	NaN	NaN	NaN	
1003009861	21905.24	17472.35	18074.19	
1003010570	0.00	0.00	0.00	
1003017443	0.00	0.00	0.00	

	Med_Sprsn_Ind	Med_Tot_HCPCS_Cds	Med_Tot_Benes	Med_Tot_Srvcs	\
NPI					
1003000134	NaN	20.0	3614.0	7835.0	
1003005315	#	NaN	NaN	NaN	
1003009861	NaN	67.0	876.0	3189.0	
1003010570	NaN	21.0	309.0	833.0	
1003017443	NaN	5.0	64.0	244.0	

	Med_Sbmtd_Chrg	Med_Mdcr_Alowd_Amt	Med_Mdcr_Pymt_Amt	\
NPI				
1003000134	1213264.0	298905.09	228078.56	
1003005315	NaN	NaN	NaN	
1003009861	725106.0	339740.57	263456.48	
1003010570	173991.0	84420.40	65891.84	
1003017443	45375.0	23956.84	18245.22	

	Med_Mdcr_Stdzd_Amt	Bene_Avg_Age	Bene_Age_LT_65_Cnt	\
NPI				
1003000134	218268.63	76	97.0	
1003005315	NaN	73	NaN	
1003009861	268694.01	73	91.0	
1003010570	66566.70	76	39.0	
1003017443	16748.76	83	NaN	

	Bene_Age_65_74_Cnt	Bene_Age_75_84_Cnt	Bene_Age_GT_84_Cnt	\
NPI				
1003000134	1623.0	1333.0	561.0	
1003005315	29.0	20.0	NaN	
1003009861	400.0	294.0	91.0	
1003010570	95.0	100.0	75.0	
1003017443	NaN	18.0	31.0	

	Bene_Feml_Cnt	Bene_Male_Cnt	Bene_Race_Wht_Cnt	\
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NPI

1003000134	1767.0	1847.0	3312.0
1003005315	44.0	27.0	NaN
1003009861	410.0	466.0	709.0
1003010570	167.0	142.0	294.0
1003017443	42.0	22.0	48.0

Bene_Race_Black_Cnt Bene_Race_API_Cnt Bene_Race_Hspnc_Cnt \

NPI

1003000134	41.0	62.0	43.0
1003005315	NaN	NaN	NaN
1003009861	35.0	20.0	80.0
1003010570	NaN	NaN	NaN
1003017443	NaN	0.0	NaN

Bene_Race_NatInd_Cnt Bene_Race_Othr_Cnt Bene_Dual_Cnt \

NPI

1003000134	0.0	156.0	195.0
1003005315	NaN	NaN	14.0
1003009861	11.0	21.0	164.0
1003010570	NaN	NaN	77.0
1003017443	0.0	NaN	NaN

Bene_Ndual_Cnt Bene_CC_BH_ADHD_OthCD_V1_Pct \

NPI

1003000134	3419.0	0.0
1003005315	57.0	NaN
1003009861	712.0	NaN
1003010570	232.0	NaN
1003017443	NaN	NaN

Bene_CC_BH_Alcohol_Drug_V1_Pct Bene_CC_BH_Tobacco_V1_Pct \

NPI

1003000134	2.0	3.0
1003005315	NaN	NaN
1003009861	13.0	21.0
1003010570	8.0	17.0
1003017443	NaN	NaN

Bene_CC_BH_Alz_NonAlzdem_V2_Pct Bene_CC_BH_Anxiety_V1_Pct \

NPI

1003000134	4.0	13.0
1003005315	NaN	35.0
1003009861	12.0	29.0
1003010570	26.0	42.0
1003017443	63.0	45.0

Bene_CC_BH_Bipolar_V1_Pct Bene_CC_BH_Mood_V2_Pct \

NPI		
1003000134	2.0	16.0
1003005315	NaN	32.0
1003009861	5.0	31.0
1003010570	9.0	46.0
1003017443	NaN	75.0

Bene_CC_BH_Depress_V1_Pct Bene_CC_BH_PD_V1_Pct \

NPI		
1003000134	14.0	1.0
1003005315	30.0	NaN
1003009861	28.0	NaN
1003010570	45.0	NaN
1003017443	75.0	NaN

Bene_CC_BH_PTSD_V1_Pct Bene_CC_BH_Schizo_OthPsy_V1_Pct \

NPI		
1003000134	0.0	1.0
1003005315	NaN	NaN
1003009861	3.0	3.0
1003010570	NaN	6.0
1003017443	NaN	23.0

Bene_CC_PH_Asthma_V2_Pct Bene_CC_PH_Afib_V2_Pct \

NPI		
1003000134	6.0	16.0
1003005315	NaN	24.0
1003009861	15.0	38.0
1003010570	17.0	35.0
1003017443	NaN	30.0

Bene_CC_PH_Cancer6_V2_Pct Bene_CC_PH_CKD_V2_Pct \

NPI		
1003000134	18.0	15.0
1003005315	17.0	25.0
1003009861	19.0	42.0
1003010570	22.0	45.0
1003017443	23.0	30.0

Bene_CC_PH_COPD_V2_Pct Bene_CC_PH_Diabetes_V2_Pct \

NPI		
1003000134	10.0	21.0
1003005315	17.0	45.0
1003009861	34.0	40.0
1003010570	37.0	42.0
1003017443	36.0	41.0

Bene_CC_PH_HF_NonIHD_V2_Pct Bene_CC_PH_Hyperlipidemia_V2_Pct \

NPI		
1003000134	11.0	72.0
1003005315	NaN	63.0
1003009861	37.0	75.0
1003010570	42.0	75.0
1003017443	44.0	75.0

Bene_CC_PH_Hypertension_V2_Pct Bene_CC_PH_IschemicHeart_V2_Pct \		
NPI		
1003000134	64.0	20.0
1003005315	75.0	23.0
1003009861	75.0	61.0
1003010570	75.0	49.0
1003017443	75.0	61.0

Bene_CC_PH_Osteoporosis_V2_Pct Bene_CC_PH_Parkinson_V2_Pct \		
NPI		
1003000134	15.0	3.0
1003005315	NaN	NaN
1003009861	13.0	2.0
1003010570	23.0	NaN
1003017443	20.0	NaN

Bene_CC_PH_Arthritis_V2_Pct Bene_CC_PH_Stroke_TIA_V2_Pct \		
NPI		
1003000134	42.0	6.0
1003005315	48.0	NaN
1003009861	50.0	22.0
1003010570	59.0	18.0
1003017443	70.0	31.0

Bene_Avg_Risk_Score LASTNAME FIRSTNAME MIDNAME BUSNAME \					
NPI					
1003000134	1.1246	NaN	NaN	NaN	NaN
1003005315	1.6177	SMITH	ADAM	BRYANT	NaN
1003009861	2.1074	NaN	NaN	NaN	NaN
1003010570	2.1762	NaN	NaN	NaN	NaN
1003017443	2.2548	NaN	NaN	NaN	NaN

GENERAL		SPECIALTY UPIN		DOB \
NPI				
1003000134	NaN	NaN	NaN	NaN
1003005315	PHYSICIAN (MD, DO)	PLASTIC SURGERY	NaN	19800519.0
1003009861	NaN	NaN	NaN	NaN
1003010570	NaN	NaN	NaN	NaN
1003017443	NaN	NaN	NaN	NaN

ADDRESS		CITY STATE		ZIP EXCLTYPE \
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NPI						
1003000134		NaN	NaN	NaN	NaN	UNK
1003005315	2136 FORD PKWY 5103	SAINT PAUL	MN	55116.0		1128a1
1003009861		NaN	NaN	NaN	NaN	UNK
1003010570		NaN	NaN	NaN	NaN	UNK
1003017443		NaN	NaN	NaN	NaN	UNK

	EXCLDATE	REINDATE	WAIVERDATE	WVRSTATE	\
NPI					
1003000134	NaN	NaN	NaN	NaN	
1003005315	20221220.0	0.0	0.0	NaN	
1003009861	NaN	NaN	NaN	NaN	
1003010570	NaN	NaN	NaN	NaN	
1003017443	NaN	NaN	NaN	NaN	

	rat_Drug_Mdcr_Alowd_Amt_Drug_Mdcr_Pymt_Amt	\
NPI		
1003000134	NaN	
1003005315	NaN	
1003009861	1.253709	
1003010570	NaN	
1003017443	NaN	

	rat_Tot_Mdcr_Alowd_Amt_Med_Sbmted_Chrg	\
NPI		
1003000134	0.246364	
1003005315	NaN	
1003009861	0.498749	
1003010570	0.485200	
1003017443	0.527974	

	rat_Drug_Tot_Benes_Tot_Benes	excluded	rand
NPI			
1003000134	0.000000	0	0.994100
1003005315	NaN	1	0.253627
1003009861	0.111872	0	0.994092
1003010570	0.000000	0	0.994872
1003017443	0.000000	0	0.991242

```
[190]: plt.figure(figsize=(12, 7))

sample_df = data.sample(n=5000, random_state=42)

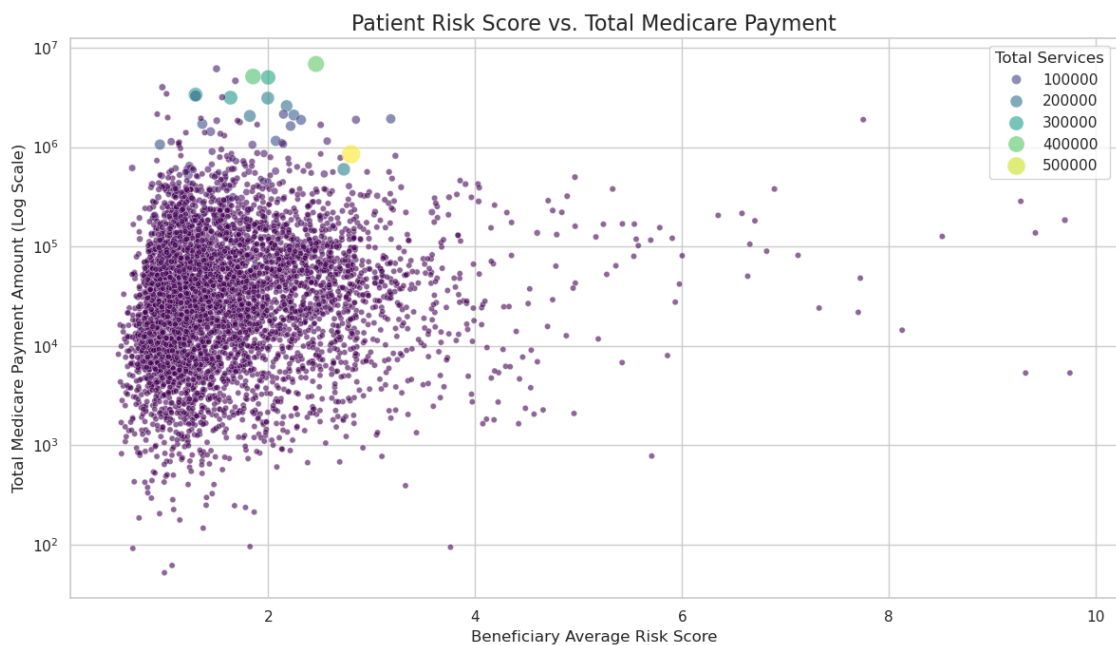
ax = sns.scatterplot(
    data=sample_df,
    x='Bene_Avg_Risk_Scre',
    y='Tot_Mdcr_Pymt_Amt',
```

```

    hue='Tot_Srvcs',
    size='Tot_Srvcs',
    sizes=(20, 200),
    palette='viridis',
    alpha=0.6
)

ax.set_yscale('log')
plt.title('Patient Risk Score vs. Total Medicare Payment', fontsize=16)
plt.xlabel('Beneficiary Average Risk Score', fontsize=12)
plt.ylabel('Total Medicare Payment Amount (Log Scale)', fontsize=12)
plt.legend(title='Total Services')
plt.tight_layout()
plt.style.use("seaborn-v0_8-dark-palette")
plt.show()

```



```

[191]: cc_cols_of_interest = [
    'Bene_Avg_Risk_Score',
    'Bene_CC_PH_Diabetes_V2_Pct',
    'Bene_CC_PH_Hypertension_V2_Pct',
    'Bene_CC_PH_IschemicHeart_V2_Pct',
    'Bene_CC_PH_CKD_V2_Pct',
    'Bene_CC_PH_COPD_V2_Pct',
    'Bene_CC_PH_HF_NonIHD_V2_Pct',
    'Bene_CC_BH_Depress_V1_Pct'
]

```

```

corr_matrix = data[cc_cols_of_interest].corr()

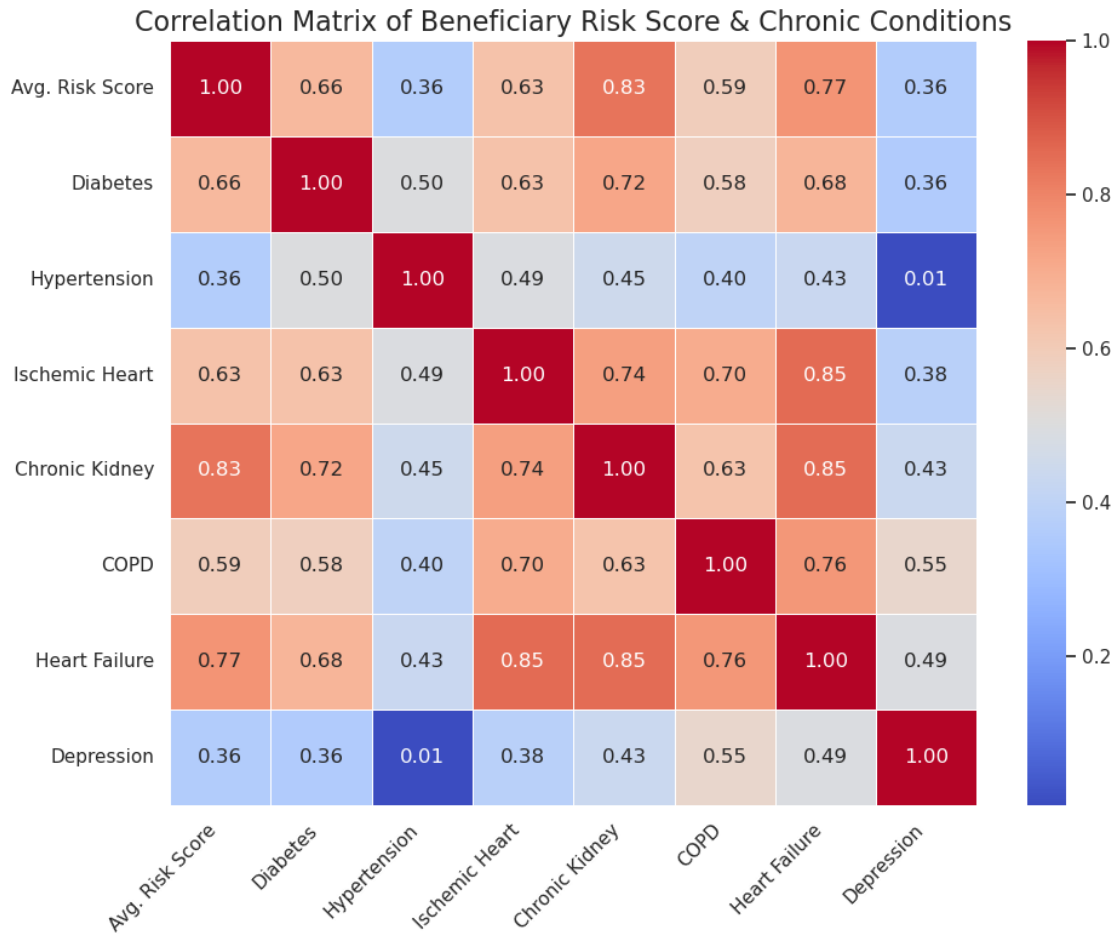
label_map = {
    'Bene_Avg_Risk_Score': 'Avg. Risk Score',
    'Bene_CC_PH_Diabetes_V2_Pct': 'Diabetes',
    'Bene_CC_PH_Hypertension_V2_Pct': 'Hypertension',
    'Bene_CC_PH_IschemicHeart_V2_Pct': 'Ischemic Heart',
    'Bene_CC_PH_CKD_V2_Pct': 'Chronic Kidney',
    'Bene_CC_PH_COPD_V2_Pct': 'COPD',
    'Bene_CC_PH_HF_NonIHD_V2_Pct': 'Heart Failure',
    'Bene_CC_BH_Depress_V1_Pct': 'Depression'
}

corr_matrix_renamed = corr_matrix.rename(columns=label_map, index=label_map)

plt.figure(figsize=(10, 8))
sns.heatmap(
    corr_matrix_renamed,
    annot=True,
    cmap='coolwarm',
    fmt=".2f",
    linewidths=.5
)

plt.title('Correlation Matrix of Beneficiary Risk Score & Chronic Conditions',
    ↪fontsize=16)
plt.xticks(rotation=45, ha='right')
plt.yticks(rotation=0)
plt.tight_layout()
plt.style.use("seaborn-v0_8-dark-palette")
plt.show()

```



1 Linear Regression/Modeling

```
[192]: df = pd.merge(data, exclusions, on='NPI',how='left')
```

```
[193]: df['excluded'].unique()
```

```
[193]: array([0, 1])
```

```
[194]: len(df)
```

```
[194]: 12496
```

```
[195]: df.columns
```

```
[195]: Index(['Rndrng_Privr_Last_Org_Name', 'Rndrng_Privr_First_Name',
        'Rndrng_Privr_MI', 'Rndrng_Privr_Crdntls', 'Rndrng_Privr_Ent_Cd',
        'Rndrng_Privr_St1', 'Rndrng_Privr_St2', 'Rndrng_Privr_City',
```

```
'Rndrng_Privr_State_Abrvtn', 'Rndrng_Privr_State_FIPS',
...
'SPECIALTY_y', 'DOB_y', 'ADDRESS_y', 'CITY_y', 'STATE_y', 'ZIP_y',
'EXCLTYPE_y', 'EXCLDATE_y', 'age_at_exclusion', 'year_excl'],
dtype='object', length=115)
```

```
[196]: df = df.copy()

# 1 if excluded (has an exclusion date), 0 otherwise
df['is_excluded'] = df['EXCLDATE_y'].notna().astype(int)

# keep rows where provider type is known
df = df[df['Rndrng_Privr_Type'].notna()]
df.head()
```

```
[196]: Rndrng_Privr_Last_Org_Name Rndrng_Privr_First_Name Rndrng_Privr_MI \
NPI
1003000134 CIBULL THOMAS L
1003005315 SMITH ADAM B
1003009861 BANNA MOUSTAFA NaN
1003010570 CHOW LING S
1003017443 BLOKAR MIRJANA NaN

Rndrng_Privr_Crdntls Rndrng_Privr_Ent_Cd Rndrng_Privr_St1 \
NPI
1003000134 M.D. I 2650 RIDGE AVE
1003005315 MD I 4977 SKYVIEW CT
1003009861 MD I 5859 W. TALAVI BLVD
1003010570 M.D. I 900 E BROADWAY AVE
1003017443 M.D. I 65 BLEECKER ST

Rndrng_Privr_St2 Rndrng_Privr_City Rndrng_Privr_State_Abrvtn \
NPI
1003000134 EVANSTON HOSPITAL EVANSTON IL
1003005315 NaN TRAVERSE CITY MI
1003009861 SUITE 100 GLENDALE AZ
1003010570 NaN BISMARCK ND
1003017443 12TH FLOOR NEW YORK NY

Rndrng_Privr_State_FIPS Rndrng_Privr_Zip5 Rndrng_Privr_RUCA \
NPI
1003000134 17 60201.0 1.0
1003005315 26 49684.0 4.0
1003009861 4 85306.0 1.0
1003010570 38 58501.0 1.0
1003017443 36 10012.0 1.0
```

	Rndrng_Prvdrr_RUCA_Desc \
NPI	
1003000134	Metropolitan area core: primary flow within an...
1003005315	Micropolitan area core: primary flow within an...
1003009861	Metropolitan area core: primary flow within an...
1003010570	Metropolitan area core: primary flow within an...
1003017443	Metropolitan area core: primary flow within an...

	Rndrng_Prvdrr_Cntry	Rndrng_Prvdrr_Type \
NPI		
1003000134	US	Pathology
1003005315	US	Plastic and Reconstructive Surgery
1003009861	US	Cardiology
1003010570	US	Internal Medicine
1003017443	US	Psychiatry

	Rndrng_Prvdrr_Mdcr_Prtcptg_Ind	Tot_HCPCS_Cds	Tot_Benes	Tot_Srvcs \
NPI				
1003000134	Y	20	3614	7835.0
1003005315	Y	51	71	533.0
1003009861	Y	68	876	3581.0
1003010570	Y	21	309	833.0
1003017443	Y	5	64	244.0

	Tot_Sbmttd_Chrg	Tot_Mdcr_Alowd_Amt	Tot_Mdcr_Pymt_Amt \
NPI			
1003000134	1213264.0	298905.09	228078.56
1003005315	250569.0	86114.40	67712.67
1003009861	788610.0	361645.81	280928.83
1003010570	173991.0	84420.40	65891.84
1003017443	45375.0	23956.84	18245.22

	Tot_Mdcr_Stdzd_Amt	Drug_Sprsn_Ind	Drug_Tot_HCPCS_Cds \
NPI			
1003000134	218268.63	NaN	0.0
1003005315	70489.71	*	NaN
1003009861	286768.20	NaN	1.0
1003010570	66566.70	NaN	0.0
1003017443	16748.76	NaN	0.0

	Drug_Tot_Benes	Drug_Tot_Srvcs	Drug_Sbmttd_Chrg \
NPI			
1003000134	0.0	0.0	0.0
1003005315	NaN	NaN	NaN
1003009861	98.0	392.0	63504.0
1003010570	0.0	0.0	0.0
1003017443	0.0	0.0	0.0

	Drug_Mdcr_Alowd_Amt	Drug_Mdcr_Pymt_Amt	Drug_Mdcr_Stdzd_Amt	\
NPI				
1003000134	0.00	0.00	0.00	
1003005315	NaN	NaN	NaN	
1003009861	21905.24	17472.35	18074.19	
1003010570	0.00	0.00	0.00	
1003017443	0.00	0.00	0.00	

	Med_Sprsn_Ind	Med_Tot_HCPCS_Cds	Med_Tot_Benes	Med_Tot_Srvcs	\
NPI					
1003000134	NaN	20.0	3614.0	7835.0	
1003005315	#	NaN	NaN	NaN	
1003009861	NaN	67.0	876.0	3189.0	
1003010570	NaN	21.0	309.0	833.0	
1003017443	NaN	5.0	64.0	244.0	

	Med_Sbmtld_Chrg	Med_Mdcr_Alowd_Amt	Med_Mdcr_Pymt_Amt	\
NPI				
1003000134	1213264.0	298905.09	228078.56	
1003005315	NaN	NaN	NaN	
1003009861	725106.0	339740.57	263456.48	
1003010570	173991.0	84420.40	65891.84	
1003017443	45375.0	23956.84	18245.22	

	Med_Mdcr_Stdzd_Amt	Bene_Avg_Age	Bene_Age_LT_65_Cnt	\
NPI				
1003000134	218268.63	76	97.0	
1003005315	NaN	73	NaN	
1003009861	268694.01	73	91.0	
1003010570	66566.70	76	39.0	
1003017443	16748.76	83	NaN	

	Bene_Age_65_74_Cnt	Bene_Age_75_84_Cnt	Bene_Age_GT_84_Cnt	\
NPI				
1003000134	1623.0	1333.0	561.0	
1003005315	29.0	20.0	NaN	
1003009861	400.0	294.0	91.0	
1003010570	95.0	100.0	75.0	
1003017443	NaN	18.0	31.0	

	Bene_Feml_Cnt	Bene_Male_Cnt	Bene_Race_Wht_Cnt	\
NPI				
1003000134	1767.0	1847.0	3312.0	
1003005315	44.0	27.0	NaN	
1003009861	410.0	466.0	709.0	
1003010570	167.0	142.0	294.0	

1003017443	42.0	22.0	48.0
------------	------	------	------

	Bene_Race_Black_Cnt	Bene_Race_API_Cnt	Bene_Race_Hspnc_Cnt \
NPI			
1003000134	41.0	62.0	43.0
1003005315	NaN	NaN	NaN
1003009861	35.0	20.0	80.0
1003010570	NaN	NaN	NaN
1003017443	NaN	0.0	NaN

	Bene_Race_NatInd_Cnt	Bene_Race_Othr_Cnt	Bene_Dual_Cnt \
NPI			
1003000134	0.0	156.0	195.0
1003005315	NaN	NaN	14.0
1003009861	11.0	21.0	164.0
1003010570	NaN	NaN	77.0
1003017443	0.0	NaN	NaN

	Bene_Ndual_Cnt	Bene_CC_BH_ADHD_OthCD_V1_Pct \
NPI		
1003000134	3419.0	0.0
1003005315	57.0	NaN
1003009861	712.0	NaN
1003010570	232.0	NaN
1003017443	NaN	NaN

	Bene_CC_BH_Alcohol_Drug_V1_Pct	Bene_CC_BH_Tobacco_V1_Pct \
NPI		
1003000134	2.0	3.0
1003005315	NaN	NaN
1003009861	13.0	21.0
1003010570	8.0	17.0
1003017443	NaN	NaN

	Bene_CC_BH_Alz_NonAlzdem_V2_Pct	Bene_CC_BH_Anxiety_V1_Pct \
NPI		
1003000134	4.0	13.0
1003005315	NaN	35.0
1003009861	12.0	29.0
1003010570	26.0	42.0
1003017443	63.0	45.0

	Bene_CC_BH_Bipolar_V1_Pct	Bene_CC_BH_Mood_V2_Pct \
NPI		
1003000134	2.0	16.0
1003005315	NaN	32.0
1003009861	5.0	31.0

1003010570	9.0	46.0
1003017443	NaN	75.0

	Bene_CC_BH_Depress_V1_Pct	Bene_CC_BH_PD_V1_Pct \
NPI		
1003000134	14.0	1.0
1003005315	30.0	NaN
1003009861	28.0	NaN
1003010570	45.0	NaN
1003017443	75.0	NaN

	Bene_CC_BH_PTSD_V1_Pct	Bene_CC_BH_Schizo_OthPsy_V1_Pct \
NPI		
1003000134	0.0	1.0
1003005315	NaN	NaN
1003009861	3.0	3.0
1003010570	NaN	6.0
1003017443	NaN	23.0

	Bene_CC_PH_Asthma_V2_Pct	Bene_CC_PH_Afib_V2_Pct \
NPI		
1003000134	6.0	16.0
1003005315	NaN	24.0
1003009861	15.0	38.0
1003010570	17.0	35.0
1003017443	NaN	30.0

	Bene_CC_PH_Cancer6_V2_Pct	Bene_CC_PH_CKD_V2_Pct \
NPI		
1003000134	18.0	15.0
1003005315	17.0	25.0
1003009861	19.0	42.0
1003010570	22.0	45.0
1003017443	23.0	30.0

	Bene_CC_PH_COPD_V2_Pct	Bene_CC_PH_Diabetes_V2_Pct \
NPI		
1003000134	10.0	21.0
1003005315	17.0	45.0
1003009861	34.0	40.0
1003010570	37.0	42.0
1003017443	36.0	41.0

	Bene_CC_PH_HF_NonIHD_V2_Pct	Bene_CC_PH_Hyperlipidemia_V2_Pct \
NPI		
1003000134	11.0	72.0
1003005315	NaN	63.0

1003009861	37.0	75.0
1003010570	42.0	75.0
1003017443	44.0	75.0

	Bene_CC_PH_Hypertension_V2_Pct	Bene_CC_PH_IschemicHeart_V2_Pct	\
NPI			
1003000134	64.0	20.0	
1003005315	75.0	23.0	
1003009861	75.0	61.0	
1003010570	75.0	49.0	
1003017443	75.0	61.0	

	Bene_CC_PH_Osteoporosis_V2_Pct	Bene_CC_PH_Parkinson_V2_Pct	\
NPI			
1003000134	15.0	3.0	
1003005315	NaN	NaN	
1003009861	13.0	2.0	
1003010570	23.0	NaN	
1003017443	20.0	NaN	

	Bene_CC_PH_Arthritis_V2_Pct	Bene_CC_PH_Stroke_TIA_V2_Pct	\
NPI			
1003000134	42.0	6.0	
1003005315	48.0	NaN	
1003009861	50.0	22.0	
1003010570	59.0	18.0	
1003017443	70.0	31.0	

	Bene_Avg_Risk_Score	LASTNAME_x	FIRSTNAME_x	MIDNAME	BUSNAME	\
NPI						
1003000134	1.1246	NaN	NaN	NaN	NaN	
1003005315	1.6177	SMITH	ADAM	BRYANT	NaN	
1003009861	2.1074	NaN	NaN	NaN	NaN	
1003010570	2.1762	NaN	NaN	NaN	NaN	
1003017443	2.2548	NaN	NaN	NaN	NaN	

	GENERAL_x	SPECIALTY_x	UPIN	DOB_x	\
NPI					
1003000134	NaN	NaN	NaN	NaN	
1003005315	PHYSICIAN (MD, DO)	PLASTIC SURGERY	NaN	19800519.0	
1003009861	NaN	NaN	NaN	NaN	
1003010570	NaN	NaN	NaN	NaN	
1003017443	NaN	NaN	NaN	NaN	

	ADDRESS_x	CITY_x	STATE_x	ZIP_x	EXCLTYPE_x	\
NPI						
1003000134	NaN	NaN	NaN	NaN	UNK	

1003005315	2136 FORD PKWY 5103	SAINT PAUL	MN	55116.0	1128a1
1003009861	NaN	NaN	NaN	NaN	UNK
1003010570	NaN	NaN	NaN	NaN	UNK
1003017443	NaN	NaN	NaN	NaN	UNK

	EXCLDATE_x	REINDATE	WAIVERDATE	WVRSTATE	\
NPI					
1003000134	NaN	NaN	NaN	NaN	
1003005315	20221220.0	0.0	0.0	NaN	
1003009861	NaN	NaN	NaN	NaN	
1003010570	NaN	NaN	NaN	NaN	
1003017443	NaN	NaN	NaN	NaN	

	rat_Drug_Mdcr_Alowd_Amt_Drug_Mdcr_Pymt_Amt	\
NPI		
1003000134	NaN	
1003005315	NaN	
1003009861	1.253709	
1003010570	NaN	
1003017443	NaN	

	rat_Tot_Mdcr_Alowd_Amt_Med_Sbmted_Chrg	\
NPI		
1003000134	0.246364	
1003005315	NaN	
1003009861	0.498749	
1003010570	0.485200	
1003017443	0.527974	

	rat_Drug_Tot_Benes_Tot_Benes	excluded	rand	LASTNAME_y	\
NPI					
1003000134	0.000000	0	0.994100	NaN	
1003005315	NaN	1	0.253627	SMITH	
1003009861	0.111872	0	0.994092	NaN	
1003010570	0.000000	0	0.994872	NaN	
1003017443	0.000000	0	0.991242	NaN	

	FIRSTNAME_y	GENERAL_y	SPECIALTY_y	DOB_y	\
NPI					
1003000134	NaN	NaN	NaN	NaT	
1003005315	ADAM	PHYSICIAN (MD, DO)	PLASTIC SURGERY	1980-05-19	
1003009861	NaN	NaN	NaN	NaT	
1003010570	NaN	NaN	NaN	NaT	
1003017443	NaN	NaN	NaN	NaT	

	ADDRESS_y	CITY_y	STATE_y	ZIP_y	\
NPI					

1003000134		NaN	NaN	NaN	NaN
1003005315	2136 FORD PKWY 5103	SAINT PAUL	MN	55116.0	
1003009861		NaN	NaN	NaN	NaN
1003010570		NaN	NaN	NaN	NaN
1003017443		NaN	NaN	NaN	NaN

	EXCLTYPE_y	EXCLDATE_y	age_at_exclusion	\
NPI				
1003000134		NaN	NaT	NaN
1003005315	Conviction - Medicare Fraud	2022-12-20		42.587269
1003009861		NaN	NaT	NaN
1003010570		NaN	NaT	NaN
1003017443		NaN	NaT	NaN

	year_excl	is_excluded
NPI		
1003000134	NaN	0
1003005315	2022.0	1
1003009861	NaN	0
1003010570	NaN	0
1003017443	NaN	0

```
[197]: by_type = (
    df.groupby('Rndrng_Prvrdr_Type')['is_excluded']
      .agg(['sum', 'count'])
      .rename(columns={'sum': 'n_excluded', 'count': 'n_total'})
      .sort_values('n_total', ascending=False)
)

print(by_type.head(20))
print(by_type.tail(20))
```

	n_excluded	n_total
Rndrng_Prvrdr_Type		
Nurse Practitioner	87	1482
Internal Medicine	116	1026
Family Practice	150	1016
Physician Assistant	23	934
Physical Therapist in Private Practice	20	605
Certified Registered Nurse Anesthetist (CRNA)	16	499
Emergency Medicine	35	478
Anesthesiology	35	427
Chiropractic	33	397
Diagnostic Radiology	8	310
Optometry	3	310
Mass Immunizer Roster Biller	0	307
Obstetrics & Gynecology	28	263
General Surgery	15	251

Psychiatry	39	242
Licensed Clinical Social Worker	17	230
Orthopedic Surgery	12	229
Cardiology	8	193
Podiatry	26	190
Psychologist, Clinical	8	179
	n_excluded	n_total
Rndrng_Privr_Type		
Gynecological Oncology	0	12
Maxillofacial Surgery	0	12
Pharmacy	0	11
Speech Language Pathologist	0	11
Hematology	0	9
Sports Medicine	1	9
Preventive Medicine	2	7
Undefined Physician type	1	6
Nuclear Medicine	1	6
Sleep Medicine	0	4
Advanced Heart Failure and Transplant Cardiology	0	4
Osteopathic Manipulative Medicine	0	4
Public Health or Welfare Agency	0	3
Addiction Medicine	2	3
Portable X-Ray Supplier	0	3
Hematopoietic Cell Transplantation and Cellular...	0	2
Neuropsychiatry	1	2
Dentist	0	1
Mammography Center	0	1
Radiation Therapy Center	0	1

```
[198]: min_n = 30

# mark small or perfectly separated types
small_or_perfect = by_type[
    (by_type['n_total'] < min_n) |
    (by_type['n_excluded'] == 0) |
    (by_type['n_excluded'] == by_type['n_total'])
].index

# create a cleaned copy
df_clean = df.copy()
df_clean = df_clean[df_clean['Rndrng_Privr_Type'].notna()].copy()

# collapse problematic types
df_clean['prov_type_collapsed'] = df_clean['Rndrng_Privr_Type'].where(
    ~df_clean['Rndrng_Privr_Type'].isin(small_or_perfect),
    other='Other'
)
```

```
print(df_clean['prov_type_collapsed'].value_counts())
```

prov_type_collapsed	
Nurse Practitioner	1482
Other	1136
Internal Medicine	1026
Family Practice	1016
Physician Assistant	934
Physical Therapist in Private Practice	605
Certified Registered Nurse Anesthetist (CRNA)	499
Emergency Medicine	478
Anesthesiology	427
Chiropractic	397
Diagnostic Radiology	310
Optometry	310
Obstetrics & Gynecology	263
General Surgery	251
Psychiatry	242
Licensed Clinical Social Worker	230
Orthopedic Surgery	229
Cardiology	193
Podiatry	190
Psychologist, Clinical	179
Ophthalmology	166
Gastroenterology	153
Neurology	147
Hospitalist	134
Dermatology	128
Pathology	117
Otolaryngology	115
Physical Medicine and Rehabilitation	113
Pulmonary Disease	102
Hematology-Oncology	100
Nephrology	98
Urology	88
Occupational Therapist in Private Practice	77
Infectious Disease	72
General Practice	68
Neurosurgery	63
Endocrinology	62
Rheumatology	46
Critical Care (Intensivists)	43
Pain Management	38
Plastic and Reconstructive Surgery	37
Radiation Oncology	35
Medical Oncology	34
Allergy/ Immunology	32

Clinical Cardiac Electrophysiology
Name: count, dtype: int64

31

```
[199]: import statsmodels.api as sm
import patsy

y, X = patsy.dmatrices('is_excluded ~ C(prov_type_collapsed)', df_clean,
    ↪return_type='dataframe')
logit_reg = sm.Logit(y, X).fit_regularized()
print(logit_reg.summary())
```

Optimization terminated successfully (Exit mode 0)
Current function value: 0.22972037051460392
Iterations: 237
Function evaluations: 238
Gradient evaluations: 237

Logit Regression Results

```
=====
Dep. Variable:          is_excluded    No. Observations:          12496
Model:                  Logit          Df Residuals:              12451
Method:                 MLE           Df Model:                  44
Date:                  Wed, 17 Dec 2025    Pseudo R-squ.:            0.07320
Time:                  22:15:43          Log-Likelihood:           -2870.6
converged:              True             LL-Null:                 -3097.3
Covariance Type:        nonrobust         LLR p-value:              2.193e-69
=====
```

```
=====
coef    std err          z      P>|z|    [0.025    0.975]
-----
Intercept
-2.7291    0.737    -3.703    0.000    -4.174    -1.284
C(prov_type_collapsed)[T.Anesthesiology]
0.3128    0.758    0.413    0.680    -1.173    1.798
C(prov_type_collapsed)[T.Cardiology]
-0.4148    0.821    -0.505    0.613    -2.024    1.194
C(prov_type_collapsed)[T.Certified Registered Nurse Anesthetist (CRNA)]
-0.6776    0.780    -0.869    0.385    -2.206    0.850
C(prov_type_collapsed)[T.Chiropractic]
0.3292    0.759    0.434    0.665    -1.159    1.817
C(prov_type_collapsed)[T.Clinical Cardiac Electrophysiology]
-0.6557    1.249    -0.525    0.600    -3.105    1.793
C(prov_type_collapsed)[T.Critical Care (Intensivists)]
-0.9906    1.245    -0.796    0.426    -3.431    1.449
C(prov_type_collapsed)[T.Dermatology]
-2.1358    1.254    -1.704    0.088    -4.593    0.322
C(prov_type_collapsed)[T.Diagnostic Radiology]
-0.8997    0.819    -1.098    0.272    -2.506    0.706
=====
```

C(prov_type_collapsed) [T.Emergency Medicine]	0.1911	0.758	0.252	0.801	-1.294	1.676
C(prov_type_collapsed) [T.Endocrinology]	-0.2511	0.945	-0.266	0.791	-2.104	1.602
C(prov_type_collapsed) [T.Family Practice]	0.9759	0.742	1.315	0.189	-0.479	2.431
C(prov_type_collapsed) [T.Gastroenterology]	-0.4683	0.846	-0.553	0.580	-2.127	1.191
C(prov_type_collapsed) [T.General Practice]	1.7822	0.785	2.270	0.023	0.243	3.321
C(prov_type_collapsed) [T.General Surgery]	-0.0254	0.784	-0.032	0.974	-1.561	1.511
C(prov_type_collapsed) [T.Hematology-Oncology]	-0.2113	0.868	-0.244	0.808	-1.912	1.490
C(prov_type_collapsed) [T.Hospitalist]	-2.1830	1.254	-1.741	0.082	-4.641	0.275
C(prov_type_collapsed) [T.Infectious Disease]	-0.8215	1.027	-0.800	0.424	-2.835	1.192
C(prov_type_collapsed) [T.Internal Medicine]	0.6693	0.744	0.900	0.368	-0.788	2.127
C(prov_type_collapsed) [T.Licensed Clinical Social Worker]	0.2008	0.779	0.258	0.797	-1.326	1.728
C(prov_type_collapsed) [T.Medical Oncology]	-0.7494	1.248	-0.601	0.548	-3.195	1.696
C(prov_type_collapsed) [T.Nephrology]	-1.8577	1.251	-1.485	0.138	-4.310	0.595
C(prov_type_collapsed) [T.Neurology]	0.2152	0.801	0.269	0.788	-1.355	1.785
C(prov_type_collapsed) [T.Neurosurgery]	0.4803	0.853	0.563	0.573	-1.191	2.152
C(prov_type_collapsed) [T.Nurse Practitioner]	-0.0456	0.745	-0.061	0.951	-1.506	1.415
C(prov_type_collapsed) [T.Obstetrics & Gynecology]	0.6021	0.764	0.788	0.430	-0.895	2.099
C(prov_type_collapsed) [T.Occupational Therapist in Private Practice]	-1.6029	1.248	-1.284	0.199	-4.049	0.843
C(prov_type_collapsed) [T.Ophthalmology]	-0.5532	0.846	-0.654	0.513	-2.212	1.105
C(prov_type_collapsed) [T.Optomety]	-1.9044	0.939	-2.028	0.043	-3.745	-0.064
C(prov_type_collapsed) [T.Orthopedic Surgery]	-0.1695	0.795	-0.213	0.831	-1.727	1.388
C(prov_type_collapsed) [T.Other]	-0.7796	0.758	-1.029	0.304	-2.265	0.706
C(prov_type_collapsed) [T.Otolaryngology]	0.1344	0.823	0.163	0.870	-1.479	1.748
C(prov_type_collapsed) [T.Pain Management]	2.5156	0.806	3.121	0.002	0.936	4.095


```

C(prov_type_collapsed)[T.Pathology]
-1.3144      1.024      -1.284      0.199      -3.321      0.692
C(prov_type_collapsed)[T.Physical Medicine and Rehabilitation]
1.4184      0.772      1.837      0.066      -0.095      2.932
C(prov_type_collapsed)[T.Physical Therapist in Private Practice]
-0.6464      0.771      -0.838      0.402      -2.158      0.865
C(prov_type_collapsed)[T.Physician Assistant]
-0.9506      0.767      -1.240      0.215      -2.453      0.552
C(prov_type_collapsed)[T.Plastic and Reconstructive Surgery]
0.6192      0.907      0.682      0.495      -1.159      2.398
C(prov_type_collapsed)[T.Podiatry]
0.8868      0.767      1.157      0.247      -0.616      2.390
C(prov_type_collapsed)[T.Psychiatry]
1.0797      0.758      1.425      0.154      -0.405      2.564
C(prov_type_collapsed)[T.Psychologist, Clinical]
-0.3367      0.821      -0.410      0.682      -1.946      1.273
C(prov_type_collapsed)[T.Pulmonary Disease]
-1.1757      1.025      -1.147      0.251      -3.184      0.832
C(prov_type_collapsed)[T.Radiation Oncology]
-0.7790      1.247      -0.625      0.532      -3.223      1.665
C(prov_type_collapsed)[T.Rheumatology]
0.8321      0.857      0.971      0.332      -0.848      2.512
C(prov_type_collapsed)[T.Urology]
-0.3154      0.897      -0.352      0.725      -2.074      1.443

```

```

=====
=====

```

```

[200]: print(
        df_clean.groupby('prov_type_collapsed')['is_excluded']
            .agg(['sum', 'count'])
    )

```

	sum	count
prov_type_collapsed		
Allergy/ Immunology	2	32
Anesthesiology	35	427
Cardiology	8	193
Certified Registered Nurse Anesthetist (CRNA)	16	499
Chiropractic	33	397
Clinical Cardiac Electrophysiology	1	31
Critical Care (Intensivists)	1	43
Dermatology	1	128
Diagnostic Radiology	8	310
Emergency Medicine	35	478
Endocrinology	3	62
Family Practice	150	1016
Gastroenterology	6	153
General Practice	19	68

General Surgery	15	251
Hematology-Oncology	5	100
Hospitalist	1	134
Infectious Disease	2	72
Internal Medicine	116	1026
Licensed Clinical Social Worker	17	230
Medical Oncology	1	34
Nephrology	1	98
Neurology	11	147
Neurosurgery	6	63
Nurse Practitioner	87	1482
Obstetrics & Gynecology	28	263
Occupational Therapist in Private Practice	1	77
Ophthalmology	6	166
Optometry	3	310
Orthopedic Surgery	12	229
Other	33	1136
Otolaryngology	8	115
Pain Management	17	38
Pathology	2	117
Physical Medicine and Rehabilitation	24	113
Physical Therapist in Private Practice	20	605
Physician Assistant	23	934
Plastic and Reconstructive Surgery	4	37
Podiatry	26	190
Psychiatry	39	242
Psychologist, Clinical	8	179
Pulmonary Disease	2	102
Radiation Oncology	1	35
Rheumatology	6	46
Urology	4	88

```
[201]: logit_model = smf.logit(
        'is_excluded ~ C(prov_type_collapsed)',
        data=df_clean
    ).fit()

print(logit_model.summary())

# odds ratios
odds_ratios = np.exp(logit_model.params)
print(odds_ratios.sort_values())
```

Optimization terminated successfully.

Current function value: 0.229720

Iterations 9

Logit Regression Results

=====

```

Dep. Variable:          is_excluded    No. Observations:          12496
Model:                  Logit          Df Residuals:              12451
Method:                 MLE           Df Model:                  44
Date:                  Wed, 17 Dec 2025    Pseudo R-squ.:            0.07320
Time:                  22:15:45          Log-Likelihood:           -2870.6
converged:              True            LL-Null:                  -3097.3
Covariance Type:        nonrobust        LLR p-value:              2.189e-69

```

```

=====
=====
coef      std err          z      P>|z|      [0.025      0.975]
-----
-----
Intercept
-2.7081      0.730      -3.708      0.000      -4.139      -1.277
C(prov_type_collapsed) [T.Anesthesiology]
0.2921      0.751      0.389      0.697      -1.180      1.765
C(prov_type_collapsed) [T.Cardiology]
-0.4329      0.815      -0.531      0.595      -2.030      1.164
C(prov_type_collapsed) [T.Certified Registered Nurse Anesthetist (CRNA)]
-0.6994      0.773      -0.904      0.366      -2.215      0.816
C(prov_type_collapsed) [T.Chiropractic]
0.3074      0.753      0.408      0.683      -1.168      1.782
C(prov_type_collapsed) [T.Clinical Cardiac Electrophysiology]
-0.6931      1.252      -0.554      0.580      -3.146      1.760
C(prov_type_collapsed) [T.Critical Care (Intensivists)]
-1.0296      1.248      -0.825      0.409      -3.475      1.416
C(prov_type_collapsed) [T.Dermatology]
-2.1361      1.241      -1.721      0.085      -4.569      0.297
C(prov_type_collapsed) [T.Diagnostic Radiology]
-0.9229      0.813      -1.135      0.257      -2.517      0.671
C(prov_type_collapsed) [T.Emergency Medicine]
0.1698      0.751      0.226      0.821      -1.302      1.642
C(prov_type_collapsed) [T.Endocrinology]
-0.2709      0.940      -0.288      0.773      -2.113      1.572
C(prov_type_collapsed) [T.Family Practice]
0.9548      0.736      1.298      0.194      -0.487      2.397
C(prov_type_collapsed) [T.Gastroenterology]
-0.4906      0.841      -0.584      0.560      -2.138      1.157
C(prov_type_collapsed) [T.General Practice]
1.7607      0.779      2.261      0.024      0.234      3.287
C(prov_type_collapsed) [T.General Surgery]
-0.0477      0.777      -0.061      0.951      -1.571      1.476
C(prov_type_collapsed) [T.Hematology-Oncology]
-0.2364      0.862      -0.274      0.784      -1.927      1.454
C(prov_type_collapsed) [T.Hospitalist]
-2.1823      1.241      -1.758      0.079      -4.615      0.251
C(prov_type_collapsed) [T.Infectious Disease]
-0.8473      1.024      -0.828      0.408      -2.853      1.159

```

C(prov_type_collapsed) [T.Internal Medicine]					
0.6482	0.737	0.880	0.379	-0.796	2.093
C(prov_type_collapsed) [T.Licensed Clinical Social Worker]					
0.1800	0.773	0.233	0.816	-1.334	1.694
C(prov_type_collapsed) [T.Medical Oncology]					
-0.7885	1.250	-0.631	0.528	-3.239	1.662
C(prov_type_collapsed) [T.Nephrology]					
-1.8667	1.242	-1.502	0.133	-4.302	0.568
C(prov_type_collapsed) [T.Neurology]					
0.1933	0.795	0.243	0.808	-1.364	1.751
C(prov_type_collapsed) [T.Neurosurgery]					
0.4568	0.847	0.539	0.590	-1.203	2.117
C(prov_type_collapsed) [T.Nurse Practitioner]					
-0.0667	0.739	-0.090	0.928	-1.514	1.381
C(prov_type_collapsed) [T.Obstetrics & Gynecology]					
0.5807	0.757	0.767	0.443	-0.903	2.065
C(prov_type_collapsed) [T.Occupational Therapist in Private Practice]					
-1.6227	1.244	-1.305	0.192	-4.060	0.815
C(prov_type_collapsed) [T.Ophthalmology]					
-0.5754	0.840	-0.685	0.494	-2.222	1.072
C(prov_type_collapsed) [T.Optomety]					
-1.9202	0.933	-2.059	0.040	-3.748	-0.092
C(prov_type_collapsed) [T.Orthopedic Surgery]					
-0.1869	0.788	-0.237	0.813	-1.732	1.358
C(prov_type_collapsed) [T.Other]					
-0.8012	0.751	-1.066	0.286	-2.274	0.671
C(prov_type_collapsed) [T.Otolaryngology]					
0.1147	0.817	0.140	0.888	-1.487	1.716
C(prov_type_collapsed) [T.Pain Management]					
2.4967	0.800	3.121	0.002	0.929	4.064
C(prov_type_collapsed) [T.Pathology]					
-1.3437	1.021	-1.316	0.188	-3.344	0.657
C(prov_type_collapsed) [T.Physical Medicine and Rehabilitation]					
1.3975	0.766	1.825	0.068	-0.103	2.898
C(prov_type_collapsed) [T.Physical Therapist in Private Practice]					
-0.6678	0.765	-0.873	0.383	-2.167	0.831
C(prov_type_collapsed) [T.Physician Assistant]					
-0.9710	0.760	-1.277	0.202	-2.461	0.519
C(prov_type_collapsed) [T.Plastic and Reconstructive Surgery]					
0.5978	0.902	0.663	0.507	-1.170	2.366
C(prov_type_collapsed) [T.Podiatry]					
0.8663	0.760	1.140	0.254	-0.624	2.356
C(prov_type_collapsed) [T.Psychiatry]					
1.0584	0.751	1.409	0.159	-0.413	2.530
C(prov_type_collapsed) [T.Psychologist, Clinical]					
-0.3542	0.815	-0.435	0.664	-1.951	1.243
C(prov_type_collapsed) [T.Pulmonary Disease]					
-1.2040	1.021	-1.179	0.239	-3.206	0.798

C(prov_type_collapsed) [T.Radiation Oncology]					
-0.8183	1.250	-0.655	0.513	-3.268	1.632
C(prov_type_collapsed) [T.Rheumatology]					
0.8109	0.851	0.952	0.341	-0.858	2.480
C(prov_type_collapsed) [T.Urology]					
-0.3365	0.892	-0.377	0.706	-2.084	1.411
=====					
=====					
Intercept					
0.066667					
C(prov_type_collapsed) [T.Hospitalist]					
0.112782					
C(prov_type_collapsed) [T.Dermatology]					
0.118110					
C(prov_type_collapsed) [T.Optometry]					
0.146580					
C(prov_type_collapsed) [T.Nephrology]					
0.154639					
C(prov_type_collapsed) [T.Occupational Therapist in Private Practice]					
0.197368					
C(prov_type_collapsed) [T.Pathology]					
0.260870					
C(prov_type_collapsed) [T.Pulmonary Disease]					
0.300000					
C(prov_type_collapsed) [T.Critical Care (Intensivists)]					
0.357143					
C(prov_type_collapsed) [T.Physician Assistant]					
0.378705					
C(prov_type_collapsed) [T.Diagnostic Radiology]					
0.397351					
C(prov_type_collapsed) [T.Infectious Disease]					
0.428571					
C(prov_type_collapsed) [T.Radiation Oncology]					
0.441176					
C(prov_type_collapsed) [T.Other]					
0.448776					
C(prov_type_collapsed) [T.Medical Oncology]					
0.454545					
C(prov_type_collapsed) [T.Certified Registered Nurse Anesthetist (CRNA)]					
0.496894					
C(prov_type_collapsed) [T.Clinical Cardiac Electrophysiology]					
0.500000					
C(prov_type_collapsed) [T.Physical Therapist in Private Practice]					
0.512821					
C(prov_type_collapsed) [T.Ophthalmology]					
0.562500					
C(prov_type_collapsed) [T.Gastroenterology]					
0.612245					

C(prov_type_collapsed) [T.Cardiology]
0.648649
C(prov_type_collapsed) [T.Psychologist, Clinical]
0.701754
C(prov_type_collapsed) [T.Urology]
0.714286
C(prov_type_collapsed) [T.Endocrinology]
0.762712
C(prov_type_collapsed) [T.Hematology-Oncology]
0.789474
C(prov_type_collapsed) [T.Orthopedic Surgery]
0.829493
C(prov_type_collapsed) [T.Nurse Practitioner]
0.935484
C(prov_type_collapsed) [T.General Surgery]
0.953390
C(prov_type_collapsed) [T.Otolaryngology]
1.121495
C(prov_type_collapsed) [T.Emergency Medicine]
1.185102
C(prov_type_collapsed) [T.Licensed Clinical Social Worker]
1.197183
C(prov_type_collapsed) [T.Neurology]
1.213235
C(prov_type_collapsed) [T.Anesthesiology]
1.339286
C(prov_type_collapsed) [T.Chiropractic]
1.359890
C(prov_type_collapsed) [T.Neurosurgery]
1.578947
C(prov_type_collapsed) [T.Obstetrics & Gynecology]
1.787234
C(prov_type_collapsed) [T.Plastic and Reconstructive Surgery]
1.818182
C(prov_type_collapsed) [T.Internal Medicine]
1.912088
C(prov_type_collapsed) [T.Rheumatology]
2.250000
C(prov_type_collapsed) [T.Podiatry]
2.378049
C(prov_type_collapsed) [T.Family Practice]
2.598152
C(prov_type_collapsed) [T.Psychiatry]
2.881773
C(prov_type_collapsed) [T.Physical Medicine and Rehabilitation]
4.044944
C(prov_type_collapsed) [T.General Practice]
5.816327

```
C(prov_type_collapsed)[T.Pain Management]
12.142857
dtype: float64
```

```
[202]: # drop the intercept
or_series = odds_ratios.drop('Intercept')

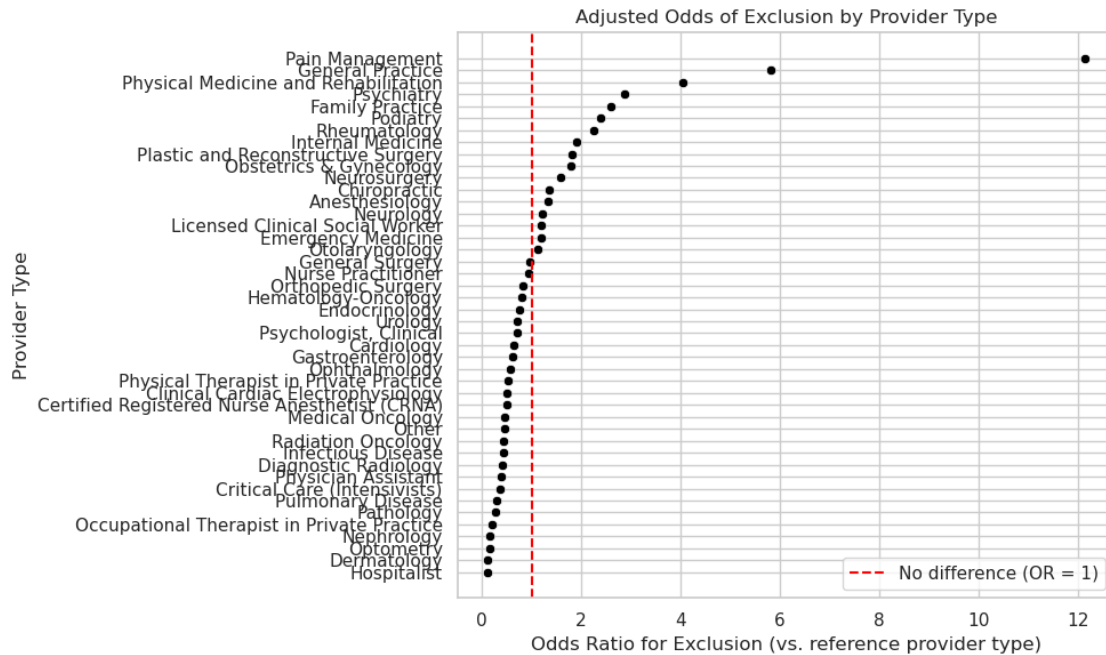
# build a DataFrame
or_df = (
    or_series
    .reset_index()
    .rename(columns={'index': 'term', 0: 'or'})
)

# extract provider type name
or_df['provider_type'] = (
    or_df['term']
    .str.replace('C(prov_type_collapsed)[T.', '', regex=False)
    .str.rstrip(']')
)

or_df = or_df.sort_values('or', ascending=False)

plt.figure(figsize=(10, 6))
sns.scatterplot(
    data=or_df,
    x='or',
    y='provider_type',
    color='black'
)

plt.axvline(1.0, color='red', linestyle='--', label='No difference (OR = 1)')
plt.xlabel('Odds Ratio for Exclusion (vs. reference provider type)')
plt.ylabel('Provider Type')
plt.title('Adjusted Odds of Exclusion by Provider Type')
plt.legend()
plt.tight_layout()
plt.show()
```



```
[203]: # Drop intercept and build DataFrame
or_series = odds_ratios.drop('Intercept')

or_df = (
    or_series.reset_index()
        .rename(columns={'index': 'term', 0: 'or'})
)

# Extract provider type names
or_df['provider_type'] = (
    or_df['term']
        .str.replace('C(prov_type_collapsed)[T.', '', regex=False)
        .str.rstrip(']')
)

# Keep top 10 by odds ratio
top10 = or_df.sort_values('or', ascending=False).head(10)

max_len = 40
top10['label'] = top10['provider_type'].str.slice(0, max_len)

# horizontal bar chart, largest at top
plt.figure(figsize=(8, 5))
sns.barplot(
```

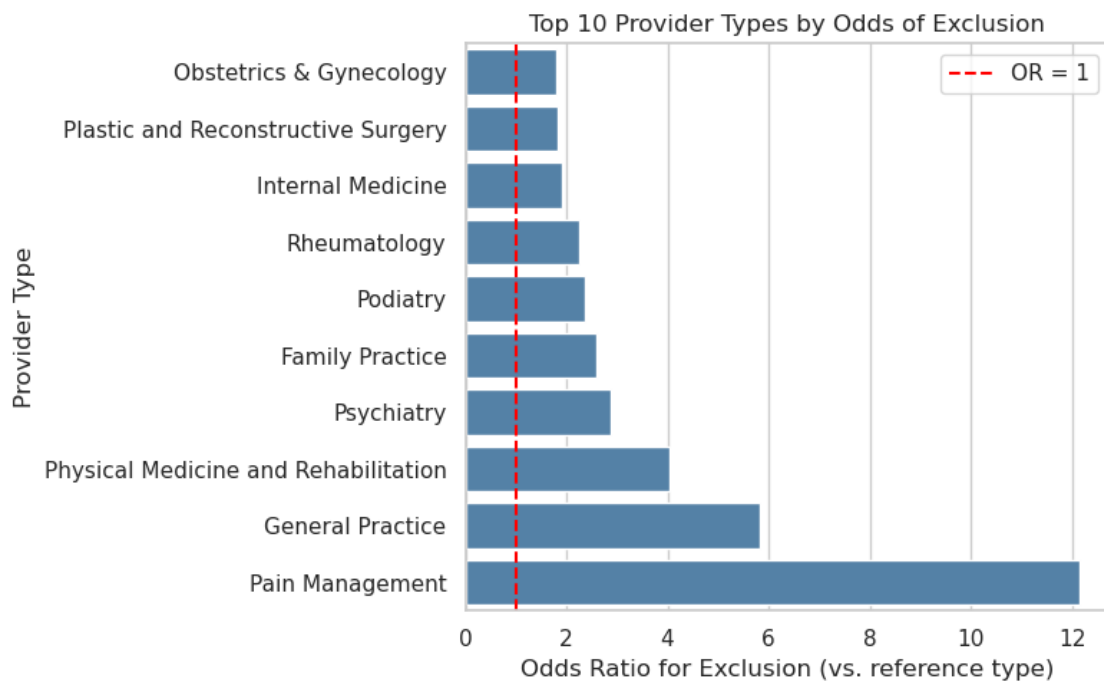


```

data=top10.sort_values('or', ascending=True), # ascending so highest at top
x='or',
y='label',
color='steelblue'
)

plt.axvline(1.0, color='red', linestyle='--', label='OR = 1')
plt.xlabel('Odds Ratio for Exclusion (vs. reference type)')
plt.ylabel('Provider Type')
plt.title('Top 10 Provider Types by Odds of Exclusion')
plt.legend()
plt.tight_layout()
plt.show()

```



```

[204]: model_df = df[[
    'excluded', # 0/1 target
    'Rndrng_Privr_Type', # provider type
    'Rndrng_Privr_State_Abrvtn', # state
    'Tot_Mdcr_Pymt_Amt' # funding
]].dropna()

# Make sure excluded is numeric 0/1
model_df['excluded'] = model_df['excluded'].astype(int)

```

```

# Log-transform funding
model_df['log_funding'] = np.log1p(model_df['Tot_Mdcr_Pymt_Amt'])

# Collapse rare provider types
type_counts = model_df['Rndrng_Privr_Type'].value_counts()
rare_types = type_counts[type_counts < 50].index
model_df['Rndrng_Privr_Type_simplified'] = model_df['Rndrng_Privr_Type'].where(
    ~model_df['Rndrng_Privr_Type'].isin(rare_types),
    other='Other'
)

# Collapse rare states
state_counts = model_df['Rndrng_Privr_State_Abrvtn'].value_counts()
rare_states = state_counts[state_counts < 50].index
model_df['Rndrng_Privr_State_simplified'] =
    model_df['Rndrng_Privr_State_Abrvtn'].where(
        ~model_df['Rndrng_Privr_State_Abrvtn'].isin(rare_states),
        other='Other'
    )

```

```

[205]: y = model_df['excluded'].astype(float)

X = model_df[[
    'Rndrng_Privr_Type_simplified',
    'Rndrng_Privr_State_simplified',
    'log_funding'
]]

# One-hot encode categoricals and ensure float dummies
X = pd.get_dummies(
    X,
    columns=['Rndrng_Privr_Type_simplified', 'Rndrng_Privr_State_simplified'],
    drop_first=True,
    dtype=float          # this avoids bool columns
)

# Drop zero-variance columns
X = X.loc[:, X.nunique() > 1]

# Add intercept and ensure float
X = sm.add_constant(X).astype(float)

print(X.dtypes.head())    # should now show all float64

logit_model = sm.Logit(y, X).fit()
print(logit_model.summary())

```

const

float64

```
log_funding float64
Rndrng_Privr_Type_simplified_Ambulatory Surgical Center float64
Rndrng_Privr_Type_simplified_Anesthesiology float64
Rndrng_Privr_Type_simplified_Audiologist float64
dtype: object
```

```
Warning: Maximum number of iterations has been exceeded.
Current function value: 0.225812
Iterations: 35
```

Logit Regression Results

```
=====
Dep. Variable:          excluded    No. Observations:          12496
Model:                  Logit      Df Residuals:              12411
Method:                 MLE       Df Model:                  84
Date:                  Wed, 17 Dec 2025    Pseudo R-squ.:          0.1041
Time:                  22:15:51    Log-Likelihood:         -2821.7
converged:              False    LL-Null:                 -3149.5
Covariance Type:        nonrobust    LLR p-value:             2.199e-89
=====
```

```
=====
coef      std err          z      P>|z|      [0.025      0.975]
-----
const
-3.9108      0.586      -6.676      0.000      -5.059      -2.763
log_funding
0.1677      0.028      5.904      0.000      0.112      0.223
Rndrng_Privr_Type_simplified_Ambulatory Surgical Center
-21.8424    2.29e+04      -0.001      0.999    -4.49e+04    4.49e+04
Rndrng_Privr_Type_simplified_Anesthesiology
0.3795      0.438      0.866      0.387      -0.480      1.239
Rndrng_Privr_Type_simplified_Audiologist
-16.1469    1836.395      -0.009      0.993    -3615.416    3583.122
Rndrng_Privr_Type_simplified_Cardiology
-0.5194      0.538      -0.965      0.334      -1.574      0.535
Rndrng_Privr_Type_simplified_Centralized Flu
-16.9419    1959.253      -0.009      0.993    -3857.007    3823.123
Rndrng_Privr_Type_simplified_Certified Registered Nurse Anesthetist (CRNA)
-0.5385      0.478      -1.128      0.259      -1.474      0.397
Rndrng_Privr_Type_simplified_Chiropractic
0.7153      0.446      1.605      0.109      -0.158      1.589
Rndrng_Privr_Type_simplified_Dermatology
-2.3001      1.081      -2.129      0.033      -4.418      -0.182
Rndrng_Privr_Type_simplified_Diagnostic Radiology
-0.9920      0.536      -1.851      0.064      -2.043      0.059
Rndrng_Privr_Type_simplified_Emergency Medicine
0.2336      0.438      0.534      0.594      -0.624      1.092
Rndrng_Privr_Type_simplified_Endocrinology
-0.1920      0.716      -0.268      0.789      -1.596      1.212
=====
```

Rndrng_Privr_Type_simplified_Family Practice	1.0588	0.410	2.581	0.010	0.255	1.863
Rndrng_Privr_Type_simplified_Gastroenterology	-0.4835	0.577	-0.837	0.402	-1.615	0.648
Rndrng_Privr_Type_simplified_General Practice	1.8904	0.492	3.844	0.000	0.927	2.854
Rndrng_Privr_Type_simplified_General Surgery	-0.0128	0.481	-0.027	0.979	-0.956	0.930
Rndrng_Privr_Type_simplified_Hematology-Oncology	-0.3763	0.610	-0.616	0.538	-1.573	0.820
Rndrng_Privr_Type_simplified_Hospitalist	-2.0079	1.081	-1.857	0.063	-4.127	0.111
Rndrng_Privr_Type_simplified_Infectious Disease	-0.8228	0.823	-1.000	0.317	-2.436	0.790
Rndrng_Privr_Type_simplified_Internal Medicine	0.6818	0.411	1.657	0.097	-0.124	1.488
Rndrng_Privr_Type_simplified_Licensed Clinical Social Worker	0.5876	0.479	1.226	0.220	-0.352	1.527
Rndrng_Privr_Type_simplified_Mass Immunizer Roster Biller	-21.2897	1.05e+04	-0.002	0.998	-2.07e+04	2.06e+04
Rndrng_Privr_Type_simplified_Nephrology	-2.0378	1.082	-1.884	0.060	-4.158	0.083
Rndrng_Privr_Type_simplified_Neurology	0.2117	0.508	0.416	0.677	-0.785	1.208
Rndrng_Privr_Type_simplified_Neurosurgery	0.4226	0.588	0.719	0.472	-0.729	1.574
Rndrng_Privr_Type_simplified_Nurse Practitioner	0.2124	0.420	0.506	0.613	-0.610	1.035
Rndrng_Privr_Type_simplified_Obstetrics & Gynecology	0.8845	0.454	1.949	0.051	-0.005	1.774
Rndrng_Privr_Type_simplified_Occupational Therapist in Private Practice	-1.3569	1.085	-1.250	0.211	-3.484	0.771
Rndrng_Privr_Type_simplified_Ophthalmology	-0.7786	0.577	-1.350	0.177	-1.909	0.352
Rndrng_Privr_Type_simplified_Optometry	-1.7136	0.707	-2.425	0.015	-3.099	-0.329
Rndrng_Privr_Type_simplified_Orthopedic Surgery	-0.2645	0.497	-0.532	0.595	-1.239	0.710
Rndrng_Privr_Type_simplified_Other	0.5608	0.416	1.349	0.177	-0.254	1.376
Rndrng_Privr_Type_simplified_Otolaryngology	0.1444	0.543	0.266	0.790	-0.920	1.209
Rndrng_Privr_Type_simplified_Pathology	-1.2742	0.819	-1.557	0.120	-2.878	0.330
Rndrng_Privr_Type_simplified_Physical Medicine and Rehabilitation	1.3806	0.462	2.986	0.003	0.474	2.287
Rndrng_Privr_Type_simplified_Physical Therapist in Private Practice	-0.4926	0.461	-1.069	0.285	-1.395	0.410

Rndrng_Privr_Type_simplified_Physician Assistant	-0.6620	0.456	-1.450	0.147	-1.557	0.233
Rndrng_Privr_Type_simplified_Podiatry	0.8173	0.452	1.807	0.071	-0.069	1.704
Rndrng_Privr_Type_simplified_Psychiatry	1.3622	0.441	3.087	0.002	0.497	2.227
Rndrng_Privr_Type_simplified_Psychologist, Clinical	-0.1819	0.541	-0.336	0.737	-1.243	0.879
Rndrng_Privr_Type_simplified_Pulmonary Disease	-1.2712	0.819	-1.553	0.120	-2.876	0.333
Rndrng_Privr_Type_simplified_Urology	-0.4419	0.650	-0.680	0.497	-1.716	0.832
Rndrng_Privr_State_simplified_AR	-0.9486	0.497	-1.910	0.056	-1.922	0.025
Rndrng_Privr_State_simplified_AZ	-0.8966	0.373	-2.401	0.016	-1.628	-0.165
Rndrng_Privr_State_simplified_CA	-0.4254	0.271	-1.570	0.116	-0.956	0.106
Rndrng_Privr_State_simplified_CO	-1.9966	0.565	-3.534	0.000	-3.104	-0.889
Rndrng_Privr_State_simplified_CT	-1.2161	0.442	-2.750	0.006	-2.083	-0.349
Rndrng_Privr_State_simplified_FL	-0.3421	0.277	-1.236	0.217	-0.885	0.200
Rndrng_Privr_State_simplified_GA	-0.9316	0.344	-2.707	0.007	-1.606	-0.257
Rndrng_Privr_State_simplified_IA	-1.1106	0.492	-2.258	0.024	-2.075	-0.147
Rndrng_Privr_State_simplified_ID	-1.2467	0.769	-1.621	0.105	-2.754	0.260
Rndrng_Privr_State_simplified_IL	-0.7329	0.310	-2.364	0.018	-1.341	-0.125
Rndrng_Privr_State_simplified_IN	-2.4720	0.633	-3.903	0.000	-3.713	-1.231
Rndrng_Privr_State_simplified_KS	-1.7287	0.640	-2.703	0.007	-2.982	-0.475
Rndrng_Privr_State_simplified_KY	-0.3963	0.359	-1.104	0.270	-1.100	0.307
Rndrng_Privr_State_simplified_LA	-0.3042	0.381	-0.799	0.424	-1.050	0.442
Rndrng_Privr_State_simplified_MA	-1.4332	0.388	-3.698	0.000	-2.193	-0.674
Rndrng_Privr_State_simplified_MD	-1.1798	0.411	-2.870	0.004	-1.986	-0.374
Rndrng_Privr_State_simplified_ME	-0.1915	0.477	-0.402	0.688	-1.126	0.743
Rndrng_Privr_State_simplified_MI	-0.4832	0.304	-1.591	0.112	-1.078	0.112

Rndrng_Privrdr_State_simplified_MN	-2.0310	0.564	-3.598	0.000	-3.137	-0.925
Rndrng_Privrdr_State_simplified_MO	-0.4958	0.347	-1.429	0.153	-1.176	0.184
Rndrng_Privrdr_State_simplified_MS	-0.1067	0.393	-0.272	0.786	-0.876	0.663
Rndrng_Privrdr_State_simplified_NC	-1.3120	0.371	-3.541	0.000	-2.038	-0.586
Rndrng_Privrdr_State_simplified_NE	-1.1256	0.575	-1.958	0.050	-2.252	0.001
Rndrng_Privrdr_State_simplified_NH	-15.6234	710.685	-0.022	0.982	-1408.541	1377.294
Rndrng_Privrdr_State_simplified_NJ	-1.2470	0.353	-3.534	0.000	-1.939	-0.555
Rndrng_Privrdr_State_simplified_NM	0.1483	0.431	0.344	0.731	-0.697	0.994
Rndrng_Privrdr_State_simplified_NV	-0.3481	0.455	-0.765	0.444	-1.240	0.544
Rndrng_Privrdr_State_simplified_NY	-0.7345	0.285	-2.580	0.010	-1.293	-0.176
Rndrng_Privrdr_State_simplified_OH	-0.3735	0.292	-1.279	0.201	-0.946	0.199
Rndrng_Privrdr_State_simplified_OK	-0.6450	0.417	-1.548	0.122	-1.462	0.172
Rndrng_Privrdr_State_simplified_OR	-0.2742	0.367	-0.748	0.455	-0.993	0.445
Rndrng_Privrdr_State_simplified_Other	-0.5573	0.319	-1.748	0.081	-1.182	0.068
Rndrng_Privrdr_State_simplified_PA	-0.6464	0.294	-2.199	0.028	-1.223	-0.070
Rndrng_Privrdr_State_simplified_PR	-0.3715	0.500	-0.744	0.457	-1.351	0.608
Rndrng_Privrdr_State_simplified_SC	-1.5645	0.487	-3.212	0.001	-2.519	-0.610
Rndrng_Privrdr_State_simplified_TN	-0.0384	0.308	-0.125	0.901	-0.641	0.565
Rndrng_Privrdr_State_simplified_TX	-0.3451	0.277	-1.248	0.212	-0.887	0.197
Rndrng_Privrdr_State_simplified_UT	-1.9977	0.761	-2.624	0.009	-3.490	-0.506
Rndrng_Privrdr_State_simplified_VA	-0.7633	0.338	-2.257	0.024	-1.426	-0.100
Rndrng_Privrdr_State_simplified_WA	-0.6014	0.339	-1.776	0.076	-1.265	0.062
Rndrng_Privrdr_State_simplified_WI	-1.3261	0.424	-3.128	0.002	-2.157	-0.495
Rndrng_Privrdr_State_simplified_WV	-0.2315	0.509	-0.455	0.649	-1.229	0.766

```
=====
/opt/conda/lib/python3.12/site-packages/statsmodels/base/model.py:607:
ConvergenceWarning: Maximum Likelihood optimization failed to converge. Check
mle_retvals
    warnings.warn("Maximum Likelihood optimization failed to "
```

```
[206]: from sklearn.metrics import roc_curve, auc

y_score = logit_model.predict(X)

fpr, tpr, _ = roc_curve(y, y_score)
roc_auc = auc(fpr, tpr)

plt.figure(figsize=(7, 7))
plt.plot(fpr, tpr, color='darkorange', lw=2,
         label=f'Logit ROC curve (AUC = {roc_auc:0.3f})')
plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--', label='Chance')

plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('ROC Curve - Provider Exclusion (Logistic Regression)')
plt.legend(loc="lower right")
plt.tight_layout()
plt.show()
```

