

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()
```

	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

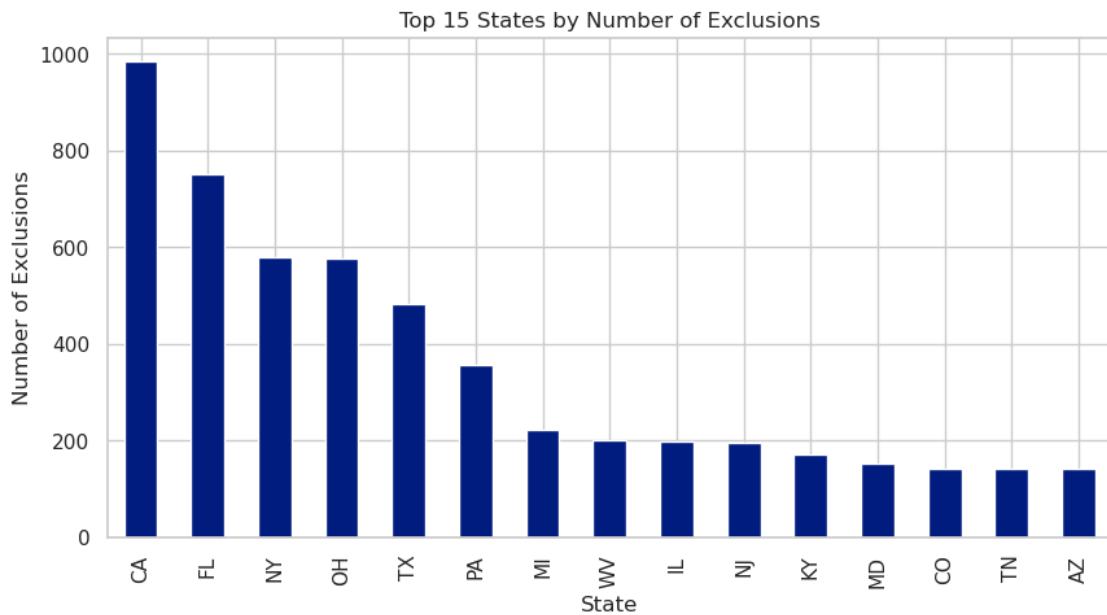
```
[180]: exclusions['SPECIALTY'].value_counts().head(15)
```

```
[180]: SPECIALTY
```

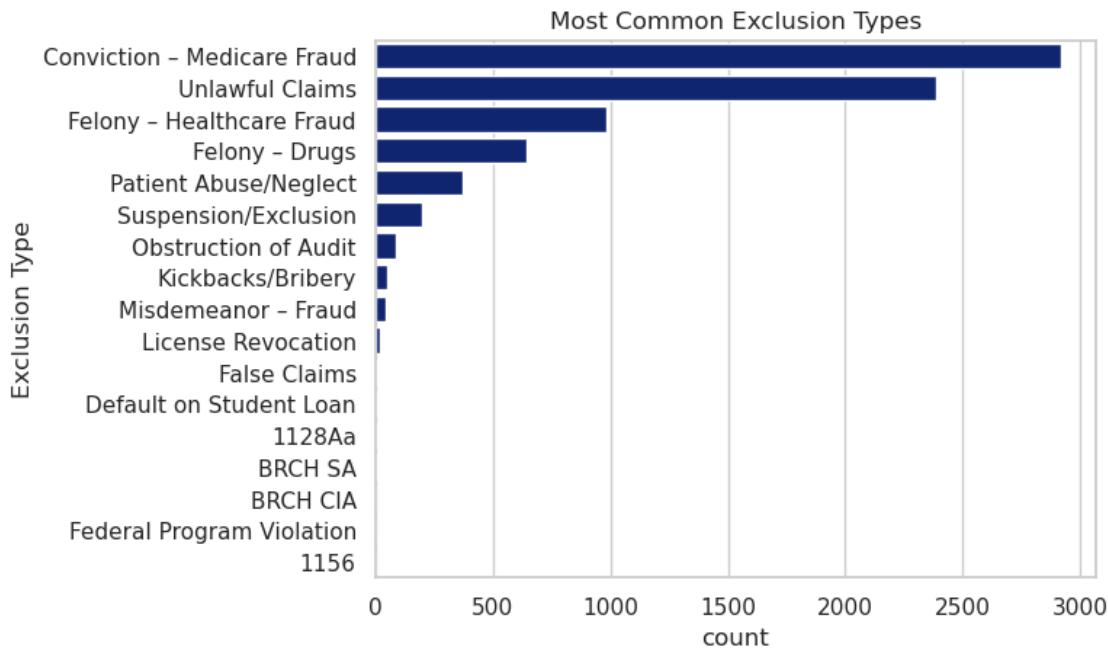
NURSE/NURSES AIDE	826
GENERAL PRACTICE	753
FAMILY PRACTICE	591
INTERNAL MEDICINE	546
CHIROPRACTIC	422
COUNSELOR	397
DENTIST	334

```
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()
```



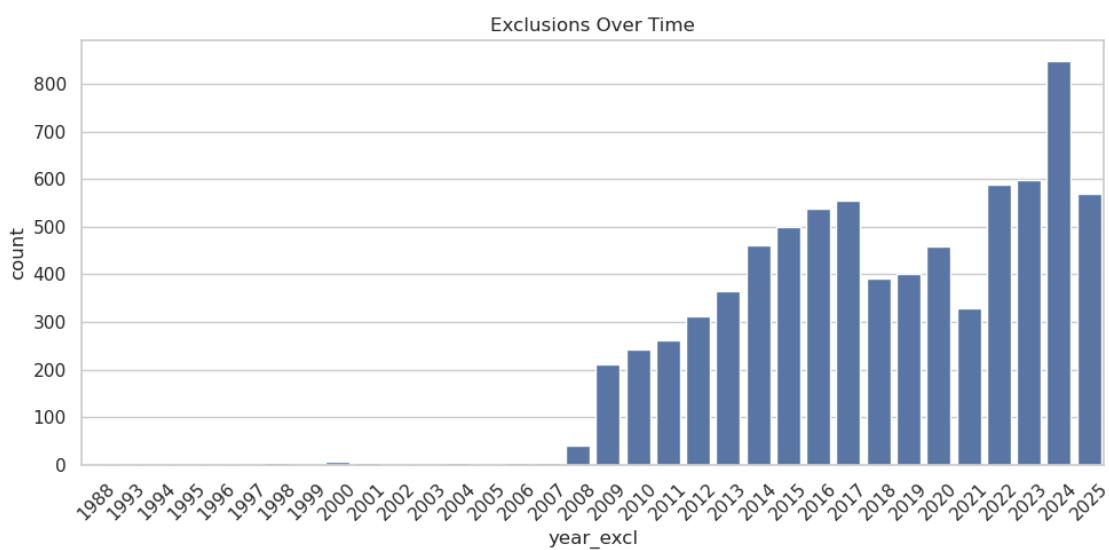
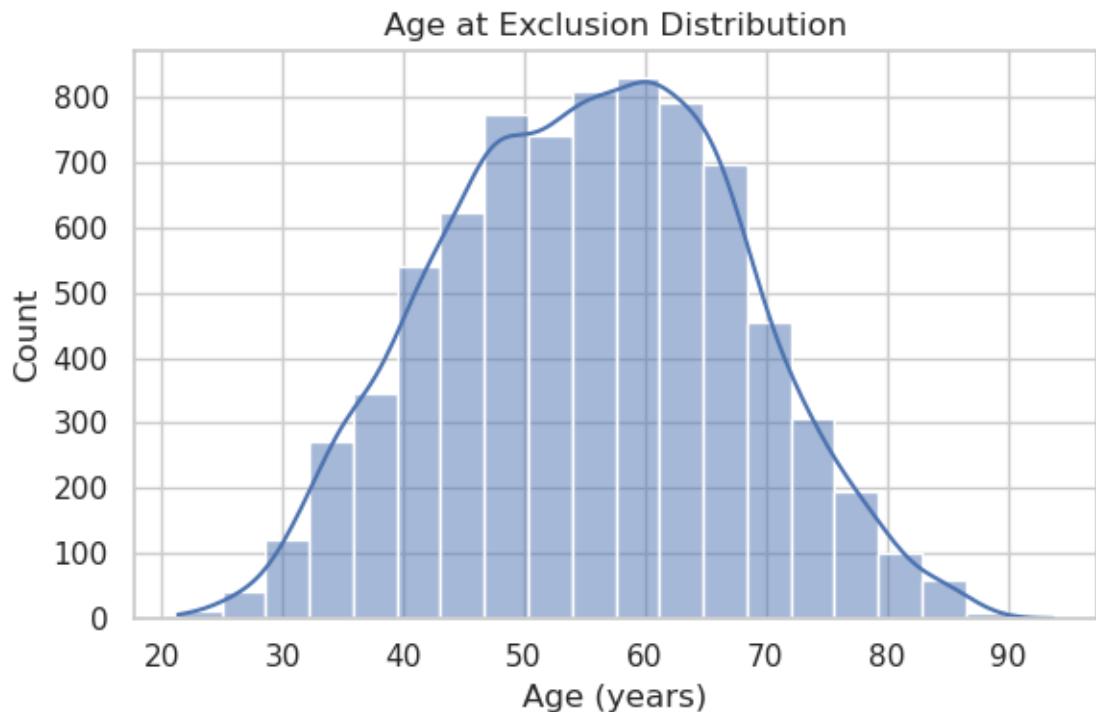
```
[182]: sns.countplot(y='EXCLTYPE', data=exclusions, order=exclusions['EXCLTYPE'].
                     value_counts().index)
plt.ylabel("Exclusion Type")
plt.title('Most Common Exclusion Types')
plt.show()
```



```
[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')

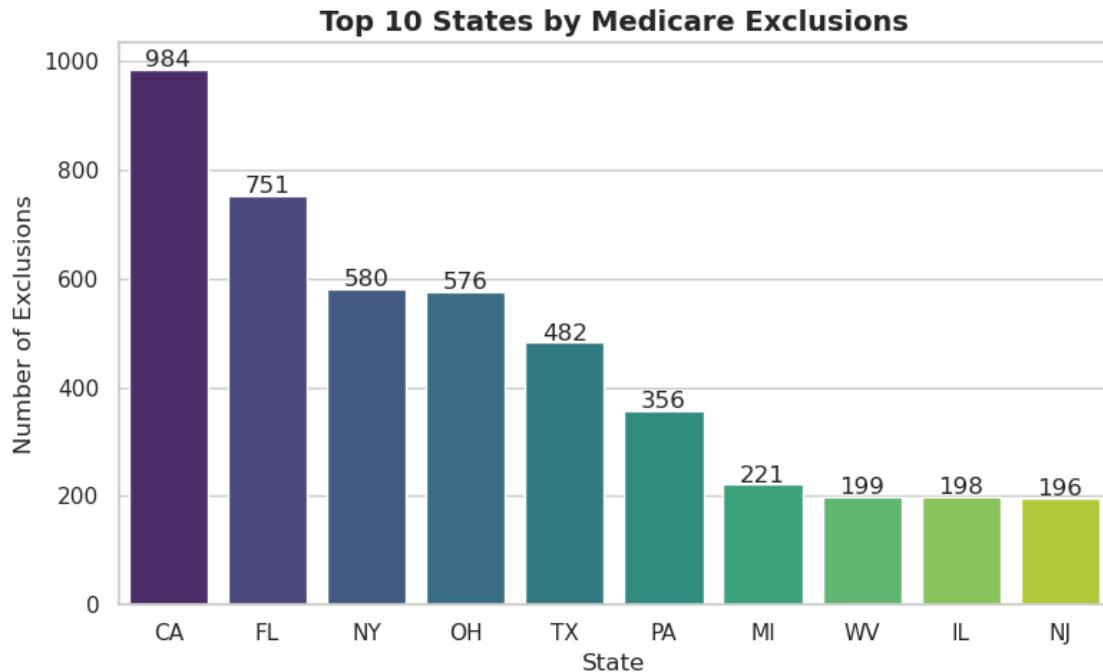
plt.tight_layout()
plt.show()

```

/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_Prvdr_Last_Org_Name Rndrng_Prvdr_First_Name Rndrng_Prvdr_MI \
NPI
1003000134 CIBULL THOMAS L
1003005315 SMITH ADAM B
1003009861 BANNA MOUSTAFA NaN
1003010570 CHOW LING S
1003017443 BLOKAR MIRJANA NaN
```

```
          Rndrng_Prvdr_Crdntls Rndrng_Prvdr_Ent_Cd Rndrng_Prvdr_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_Prvdr_St2 Rndrng_Prvdr_City Rndrng_Prvdr_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_Prvdr_State_FIPS	Rndrng_Prvdr_Zip5	Rndrng_Prvdr_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_Prvdr_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_Prvdr_Cntry	Rndrng_Prvdr_Type	\
NPI			
1003000134	US	Pathology	
1003005315	US	Plastic and Reconstructive Surgery	
1003009861	US	Cardiology	
1003010570	US	Internal Medicine	
1003017443	US	Psychiatry	

	Rndrng_Prvdr_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_Sbmtd_Chrg	TotMdcr_Alowd_Amt	Tot_Mdcr_Pynt_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_Sbmtd_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
1003005315	#	NaN	NaN
1003009861	NaN	67.0	876.0
1003010570	NaN	21.0	309.0
1003017443	NaN	5.0	64.0
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
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
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
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
NPI			
1003000134	3419.0	0.0	
1003005315	57.0	NaN	
1003009861	712.0	NaN	
1003010570	232.0	NaN	
1003017443	NaN	NaN	
NPI			
1003000134	2.0	3.0	
1003005315	NaN	NaN	
1003009861	13.0	21.0	
1003010570	8.0	17.0	
1003017443	NaN	NaN	
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_0thPsy_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
						\
NPI		ADDRESS		CITY STATE	ZIP EXCLTYPE	\
1003000134		NaN		NaN	NaN	UNK
1003005315	2136 FORD PKWY	5103 SAINT PAUL	MN	55116.0	1128a1	
1003009861		NaN		NaN	NaN	UNK
1003010570		NaN		NaN	NaN	UNK
1003017443		NaN		NaN	NaN	UNK
						\
NPI		EXCLDATE	REINDATE	WAIVERDATE	WVRSTATE	\
1003000134		NaN		NaN	NaN	
1003005315	20221220.0	0.0		0.0	NaN	
1003009861		NaN		NaN	NaN	
1003010570		NaN		NaN	NaN	
1003017443		NaN		NaN	NaN	
						\
NPI		rat_Drug_Mdcr_Alowd_Amt_Drug_Mdcr_Pymt_Amt				\
1003000134				NaN		
1003005315				NaN		
1003009861				1.253709		
1003010570					NaN	
1003017443					NaN	
						\
NPI		rat_Tot_Mdcr_Alowd_Amt_Med_Sbmtd_Chrg				\
1003000134				0.246364		
1003005315				NaN		
1003009861				0.498749		
1003010570				0.485200		
1003017443				0.527974		
						\
NPI		rat_Drug_Tot_Benes_Tot_Benes	excluded		rand	
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_Prvdr_Last_Org_Name          0
Rndrng_Prvdr_First_Name            649
Rndrng_Prvdr_MI                  4210
Rndrng_Prvdr_Crdntls            1302
Rndrng_Prvdr_Ent_Cd              0
...  

rat_Drug_Mdcr_Alowd_Amt_Drug_Mdcr_Pymt_Amt  9475
rat_Tot_Mdcr_Alowd_Amt_Med_Sbmtd_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_Prvdr_Last_Org_Name to rand
dtypes: float64(70), int64(5), object(27)
memory usage: 9.8+ MB
None

==== Sample Rows ====
    Rndrng_Prvdr_Last_Org_Name Rndrng_Prvdr_First_Name Rndrng_Prvdr_MI \
NPI
1003000134                CIBULL             THOMAS        L
1003005315                 SMITH             ADAM         B
1003009861                 BANNA            MOUSTAFA     NaN
1003010570                 CHOW              LING         S
1003017443                 BLOKAR            MIRJANA     NaN

    Rndrng_Prvdr_Crdntls Rndrng_Prvdr_Ent_Cd      Rndrng_Prvdr_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_Prvdr_St2 Rndrng_Prvdr_City Rndrng_Prvdr_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_Prvdr_State_FIPS  Rndrng_Prvdr_Zip5  Rndrng_Prvdr_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_Prvdr_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_Prvdr_Cntry		Rndrng_Prvdr_Type	\	
NPI						
1003000134	US	Pathology				
1003005315	US	Plastic and Reconstructive Surgery				
1003009861	US	Cardiology				
1003010570	US	Internal Medicine				
1003017443	US	Psychiatry				
		Rndrng_Prvdr_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_Sbmtd_Chrg	TotMdcr_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_Sbmtd_Chrg		\

NPI	Drug_Mdcr_Alowd_Amt	Drug_Mdcr_Pyamt_Amt	Drug_Mdcr_Stdzd_Amt	\
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	

NPI	Med_Sprsn_Ind	Med_Tot_HCPCS_Cds	Med_Tot_Benes	Med_Tot_Srvcs	\
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	

NPI	Med_Sbmtd_Chrg	Med_Mdcr_Alowd_Amt	Med_Mdcr_Pyamt_Amt	\
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	

NPI	Med_Mdcr_Stdzd_Amt	Bene_Avg_Age	Bene_Age_LT_65_Cnt	\
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	

NPI	Bene_Age_65_74_Cnt	Bene_Age_75_84_Cnt	Bene_Age_GT_84_Cnt	\
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	Bene_Race_Black_Cnt	Bene_Race_API_Cnt	Bene_Race_Hspnc_Cnt	\
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	

NPI	Bene_Race_NatInd_Cnt	Bene_Race_Othr_Cnt	Bene_Dual_Cnt	\
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	

NPI	Bene_Ndual_Cnt	Bene_CC_BH_ADHD_0thCD_V1_Pct	\
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

NPI	Bene_CC_BH_Alcohol_Drug_V1_Pct	Bene_CC_BH_Tobacco_V1_Pct	\
1003000134	3419.0	0.0	
1003005315	57.0	NaN	
1003009861	712.0	NaN	
1003010570	232.0	NaN	
1003017443	NaN	NaN	

NPI	Bene_CC_BH_Alz_NonAlzdem_V2_Pct	Bene_CC_BH_Anxiety_V1_Pct	\
1003000134	2.0	3.0	
1003005315	NaN	NaN	
1003009861	13.0	21.0	
1003010570	8.0	17.0	
1003017443	NaN	NaN	

NPI	Bene_CC_BH_Bipolar_V1_Pct	Bene_CC_BH_Mood_V2_Pct	\
1003000134	4.0	13.0	
1003005315	NaN	35.0	
1003009861	12.0	29.0	
1003010570	26.0	42.0	
1003017443	63.0	45.0	

NPI	Bene_CC_BH_Depress_V1_Pct	Bene_CC_BH_PD_V1_Pct	\
1003000134	2.0	16.0	
1003005315	NaN	32.0	
1003009861	5.0	31.0	
1003010570	9.0	46.0	
1003017443	NaN	75.0	

NPI	Bene_CC_BH_PTSD_V1_Pct	Bene_CC_BH_Schizo_0thPsy_V1_Pct	\
1003000134	14.0	1.0	
1003005315	30.0	NaN	
1003009861	28.0	NaN	
1003010570	45.0	NaN	
1003017443	75.0	NaN	

NPI	Bene_CC_PH_Asthma_V2_Pct	Bene_CC_PH_Afib_V2_Pct	\
1003000134	0.0	1.0	
1003005315	NaN	NaN	
1003009861	3.0	3.0	
1003010570	NaN	6.0	
1003017443	NaN	23.0	

NPI	Bene_CC_PH_Cancer6_V2_Pct	Bene_CC_PH_CKD_V2_Pct	\
1003000134	6.0	16.0	
1003005315	NaN	24.0	
1003009861	15.0	38.0	
1003010570	17.0	35.0	
1003017443	NaN	30.0	

NPI	Bene_CC_PH_COPD_V2_Pct	Bene_CC_PH_Diabetes_V2_Pct	\
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	

NPI	Bene_CC_PH_HF_NonIHD_V2_Pct	Bene_CC_PH_Hyperlipidemia_V2_Pct	\
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	

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

rat_Tot_Mdcr_Alowd_Amt_Med_Sbmtd_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_Pyamt_Amt',

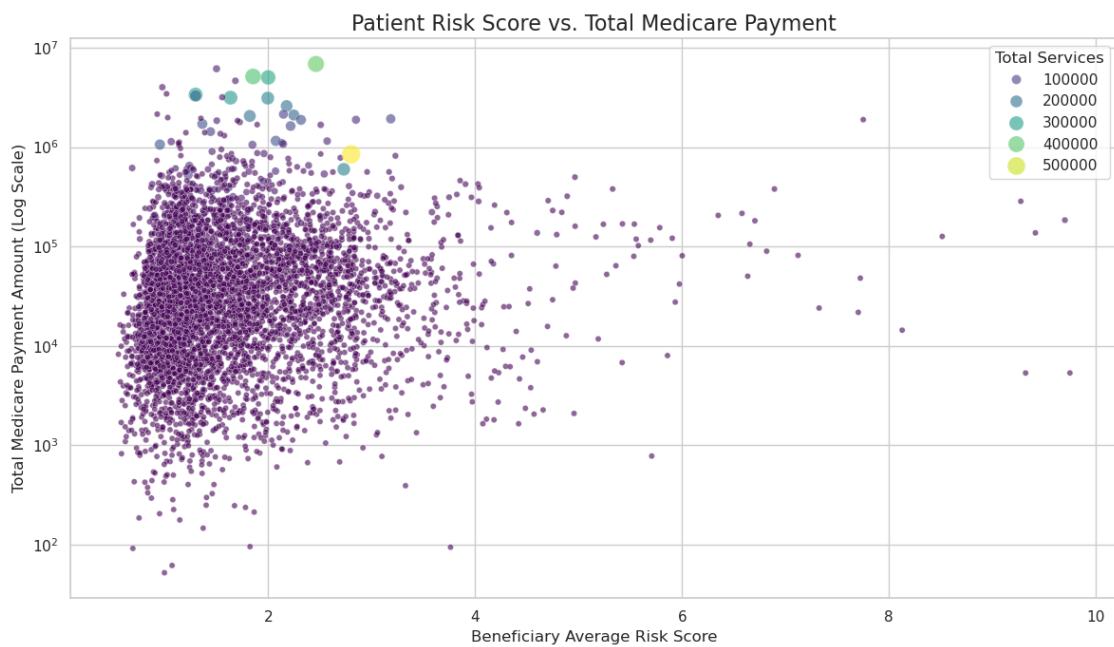
```

```

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_Scre',
    '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_Scre': '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'
}

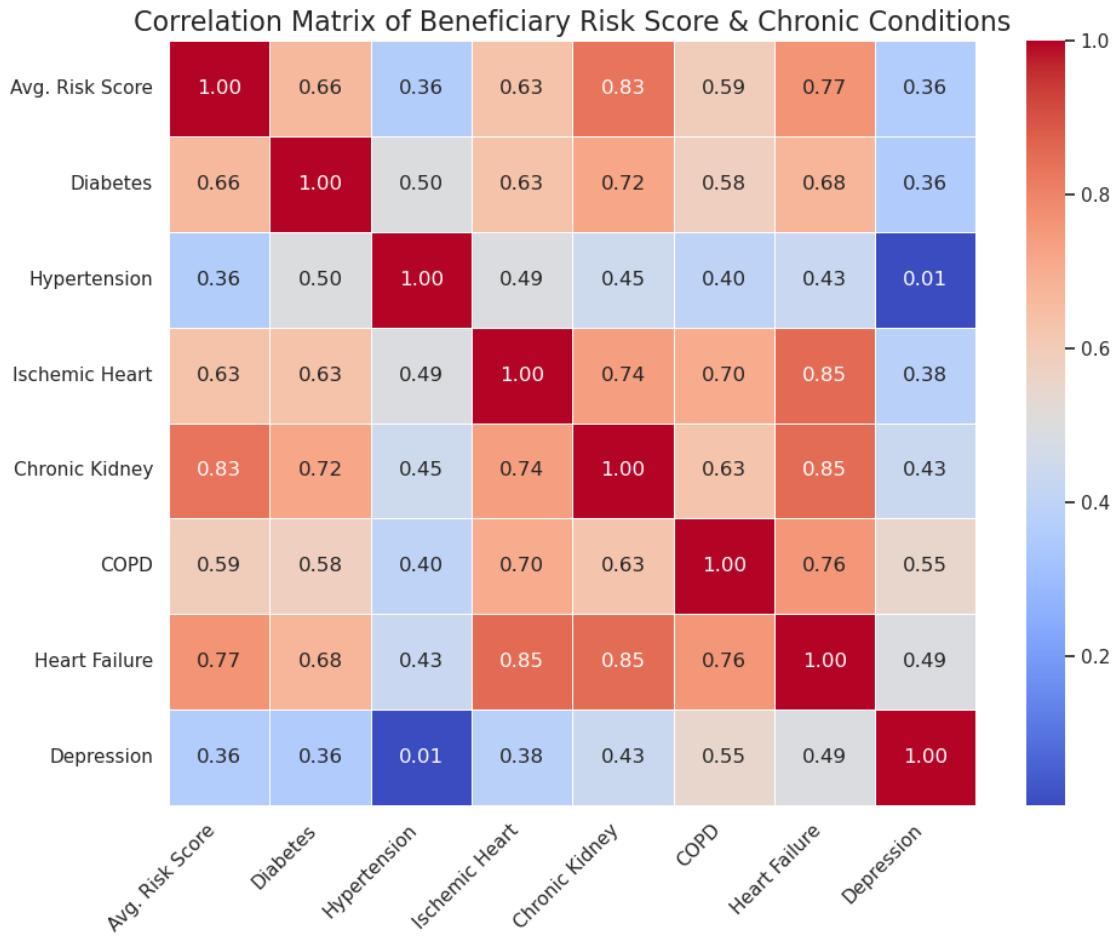
corr_matrix_renamed = corr_matrix.rename(columns=label_map, index=label_map)

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_Prvdr_Last_Org_Name', 'Rndrng_Prvdr_First_Name',
       'Rndrng_Prvdr_MI', 'Rndrng_Prvdr_Crdntls', 'Rndrng_Prvdr_Ent_Cd',
       'Rndrng_Prvdr_St1', 'Rndrng_Prvdr_St2', 'Rndrng_Prvdr_City',
```

```
'Rndrng_Prvdr_State_Abrvtn', 'Rndrng_Prvdr_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_Prvdr_Type'].notna()]
df.head()
```

	Rndrng_Prvdr_Last_Org_Name	Rndrng_Prvdr_First_Name	Rndrng_Prvdr_MI
NPI			
1003000134	CIBULL	THOMAS	L
1003005315	SMITH	ADAM	B
1003009861	BANNA	MOUSTAFA	NaN
1003010570	CHOW	LING	S
1003017443	BLOKAR	MIRJANA	NaN

	Rndrng_Prvdr_Crdntls	Rndrng_Prvdr_Ent_Cd	Rndrng_Prvdr_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_Prvdr_St2	Rndrng_Prvdr_City	Rndrng_Prvdr_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_Prvdr_State_FIPS	Rndrng_Prvdr_Zip5	Rndrng_Prvdr_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_Prvdr_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_Prvdr_Cntry			Rndrng_Prvdr_Type \		
NPI					
1003000134	US		Pathology		
1003005315	US	Plastic and Reconstructive Surgery			
1003009861	US		Cardiology		
1003010570	US		Internal Medicine		
1003017443	US		Psychiatry		

Rndrng_Prvdr_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_Sbmtd_Chrg Tot_Mdcr_Alowd_Amt Tot_Mdcr_Pynt_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_Sbmtd_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_Pyamt_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_Pyamt_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	
NPI	Bene_Race_Black_Cnt	Bene_Race_API_Cnt	Bene_Race_Hspnc_Cnt	\
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	
NPI	Bene_Race_NatInd_Cnt	Bene_Race_Othr_Cnt	Bene_Dual_Cnt	\
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	
NPI	Bene_Ndual_Cnt	Bene_CC_BH_ADHD_0thCD_V1_Pct	\	
1003000134	3419.0	0.0		
1003005315	57.0	NaN		
1003009861	712.0	NaN		
1003010570	232.0	NaN		
1003017443	NaN	NaN		
NPI	Bene_CC_BH_Alcohol_Drug_V1_Pct	Bene_CC_BH_Tobacco_V1_Pct	\	
1003000134	2.0	3.0		
1003005315	NaN	NaN		
1003009861	13.0	21.0		
1003010570	8.0	17.0		
1003017443	NaN	NaN		
NPI	Bene_CC_BH_Alz_NonAlzdem_V2_Pct	Bene_CC_BH_Anxiety_V1_Pct	\	
1003000134	4.0	13.0		
1003005315	NaN	35.0		
1003009861	12.0	29.0		
1003010570	26.0	42.0		
1003017443	63.0	45.0		
NPI	Bene_CC_BH_Bipolar_V1_Pct	Bene_CC_BH_Mood_V2_Pct	\	
1003000134	2.0	16.0		
1003005315	NaN	32.0		
1003009861	5.0	31.0		

1003010570	9.0	46.0
1003017443	NaN	75.0
NPI		
1003000134	14.0	1.0
1003005315	30.0	NaN
1003009861	28.0	NaN
1003010570	45.0	NaN
1003017443	75.0	NaN
NPI		
1003000134	0.0	1.0
1003005315	NaN	NaN
1003009861	3.0	3.0
1003010570	NaN	6.0
1003017443	NaN	23.0
NPI		
1003000134	6.0	16.0
1003005315	NaN	24.0
1003009861	15.0	38.0
1003010570	17.0	35.0
1003017443	NaN	30.0
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
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
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_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	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	UNK
1003010570		NaN	NaN	NaN	UNK
1003017443		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_Pynt_Amt \					
NPI					
1003000134			NaN		
1003005315			NaN		
1003009861			1.253709		
1003010570			NaN		
1003017443			NaN		
rat_Tot_Mdcr_Alowd_Amt_Med_Sbmtd_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_Prvdr_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))

```

Rndrng_Prvdr_Type	n_excluded	n_total
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

		n_excluded	n_total
Psychiatry	39	242	
Licensed Clinical Social Worker	17	230	
Orthopedic Surgery	12	229	
Cardiology	8	193	
Podiatry	26	190	
Psychologist, Clinical	8	179	
Rndrng_Prvdr_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_Prvdr_Type'].notna()].copy()

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

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

prov_typeCollapsed	Count
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

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

```
y, X = patsy.dmatrices('is_excluded ~ C(prov_typeCollapsed)', 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_typeCollapsed) [T.Anesthesiology]      0.3128      0.758     0.413      0.680      -1.173      1.798
C(prov_typeCollapsed) [T.Cardiology]          -0.4148      0.821     -0.505      0.613      -2.024      1.194
C(prov_typeCollapsed) [T.Certified Registered Nurse Anesthetist (CRNA)]      -0.6776      0.780     -0.869      0.385      -2.206      0.850
C(prov_typeCollapsed) [T.Chiropractic]          0.3292      0.759     0.434      0.665      -1.159      1.817
C(prov_typeCollapsed) [T.Clinical Cardiac Electrophysiology]      -0.6557      1.249     -0.525      0.600      -3.105      1.793
C(prov_typeCollapsed) [T.Critical Care (Intensivists)]      -0.9906      1.245     -0.796      0.426      -3.431      1.449
C(prov_typeCollapsed) [T.Dermatology]          -2.1358      1.254     -1.704      0.088      -4.593      0.322
C(prov_typeCollapsed) [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.Optometry]						
-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		
<hr/>					
<hr/>					
coef	std err	z	P> z	[0.025	0.975]
<hr/>					
<hr/>					
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

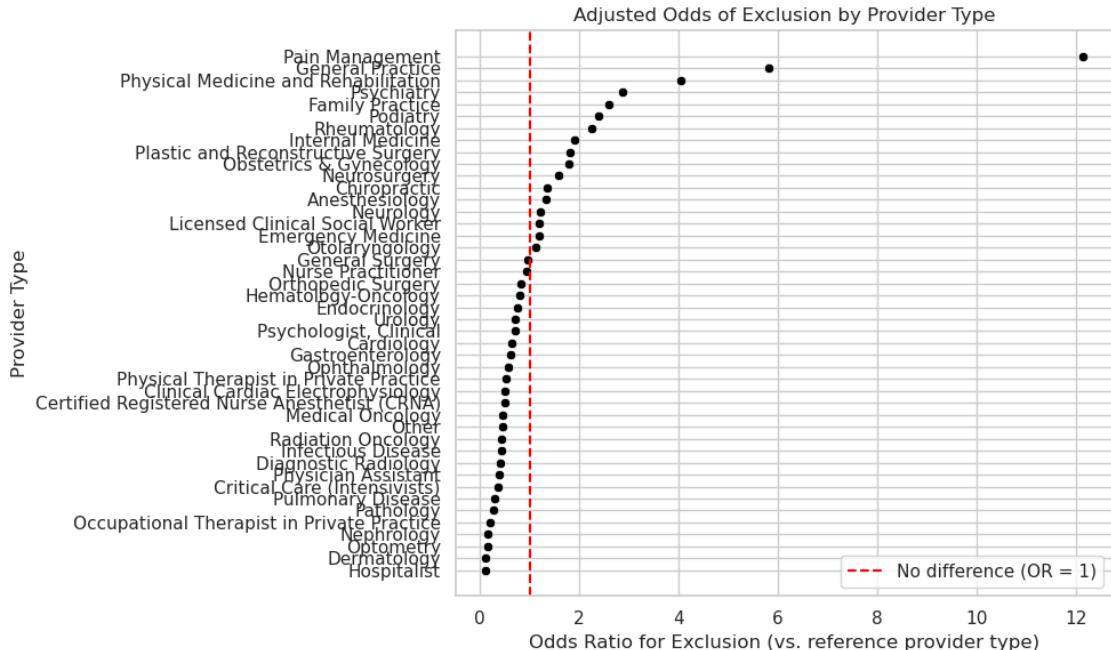
C(prov_type_collapsed) [T.Internal Medicine]						
0.6482	0.737	0.880	0.379	-0.796	2.093	
C(prov_type_collapsed) [T.Licensed Clinical Social Worker]						
0.1800	0.773	0.233	0.816	-1.334	1.694	
C(prov_type_collapsed) [T.Medical Oncology]						
-0.7885	1.250	-0.631	0.528	-3.239	1.662	
C(prov_type_collapsed) [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.Optometry]						
-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_typeCollapsed) [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_typeCollapsed)[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(

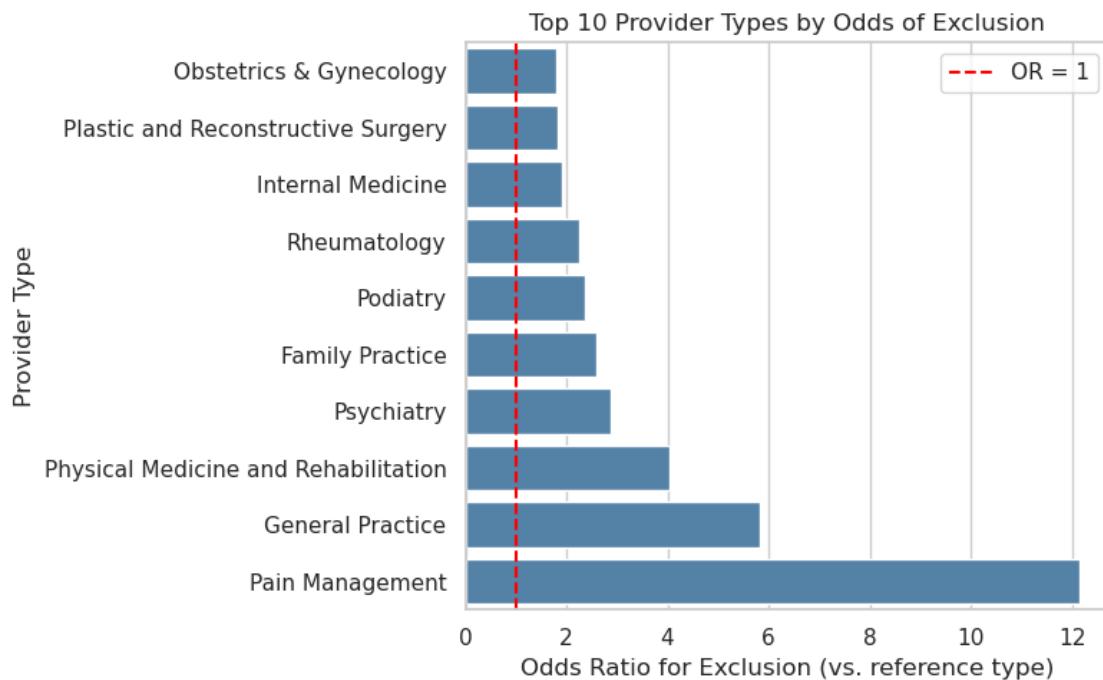
```

```

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_Prvdr_Type',             # provider type
    'Rndrng_Prvdr_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_Prvdr_Type'].value_counts()
rare_types = type_counts[type_counts < 50].index
model_df['Rndrng_Prvdr_Type_simplified'] = model_df['Rndrng_Prvdr_Type'].where(
    ~model_df['Rndrng_Prvdr_Type'].isin(rare_types),
    other='Other'
)

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

```

```

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

X = model_df[['Rndrng_Prvdr_Type_simplified',
               'Rndrng_Prvdr_State_simplified',
               'log_funding']]
              

# One-hot encode categoricals and ensure float dummies
X = pd.get_dummies(
    X,
    columns=['Rndrng_Prvdr_Type_simplified', 'Rndrng_Prvdr_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_Prvdr_Type_simplified_Ambulatory Surgical Center   float64
Rndrng_Prvdr_Type_simplified_Anesthesiology           float64
Rndrng_Prvdr_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_Prvdr_Type_simplified_Ambulatory Surgical Center
-21.8424    2.29e+04    -0.001     0.999     -4.49e+04    4.49e+04
Rndrng_Prvdr_Type_simplified_Anesthesiology
0.3795      0.438      0.866     0.387      -0.480      1.239
Rndrng_Prvdr_Type_simplified_Audiologist
-16.1469    1836.395    -0.009     0.993    -3615.416    3583.122
Rndrng_Prvdr_Type_simplified_Cardiology
-0.5194      0.538     -0.965     0.334      -1.574      0.535
Rndrng_Prvdr_Type_simplified_Centralized Flu
-16.9419    1959.253    -0.009     0.993    -3857.007    3823.123
Rndrng_Prvdr_Type_simplified_Certified Registered Nurse Anesthetist (CRNA)
-0.5385      0.478     -1.128     0.259      -1.474      0.397
Rndrng_Prvdr_Type_simplified_Chiropractic
0.7153      0.446      1.605     0.109      -0.158      1.589
Rndrng_Prvdr_Type_simplified_Dermatology
-2.3001      1.081     -2.129     0.033      -4.418     -0.182
Rndrng_Prvdr_Type_simplified_Diagnostic Radiology
-0.9920      0.536     -1.851     0.064      -2.043      0.059
Rndrng_Prvdr_Type_simplified_Emergency Medicine
0.2336      0.438      0.534     0.594      -0.624      1.092
Rndrng_Prvdr_Type_simplified_Endocrinology
-0.1920      0.716     -0.268     0.789      -1.596      1.212

```

Rndrng_Prvdr_Type_simplified_Family Practice					
1.0588	0.410	2.581	0.010	0.255	1.863
Rndrng_Prvdr_Type_simplified_Gastroenterology					
-0.4835	0.577	-0.837	0.402	-1.615	0.648
Rndrng_Prvdr_Type_simplified_General Practice					
1.8904	0.492	3.844	0.000	0.927	2.854
Rndrng_Prvdr_Type_simplified_General Surgery					
-0.0128	0.481	-0.027	0.979	-0.956	0.930
Rndrng_Prvdr_Type_simplified_Hematology-Oncology					
-0.3763	0.610	-0.616	0.538	-1.573	0.820
Rndrng_Prvdr_Type_simplified_Hospitalist					
-2.0079	1.081	-1.857	0.063	-4.127	0.111
Rndrng_Prvdr_Type_simplified_Infectious Disease					
-0.8228	0.823	-1.000	0.317	-2.436	0.790
Rndrng_Prvdr_Type_simplified_Internal Medicine					
0.6818	0.411	1.657	0.097	-0.124	1.488
Rndrng_Prvdr_Type_simplified_Licensed Clinical Social Worker					
0.5876	0.479	1.226	0.220	-0.352	1.527
Rndrng_Prvdr_Type_simplified_Mass Immunizer Roster Biller					
-21.2897	1.05e+04	-0.002	0.998	-2.07e+04	2.06e+04
Rndrng_Prvdr_Type_simplified_Nephrology					
-2.0378	1.082	-1.884	0.060	-4.158	0.083
Rndrng_Prvdr_Type_simplified_Neurology					
0.2117	0.508	0.416	0.677	-0.785	1.208
Rndrng_Prvdr_Type_simplified_Neurosurgery					
0.4226	0.588	0.719	0.472	-0.729	1.574
Rndrng_Prvdr_Type_simplified_Nurse Practitioner					
0.2124	0.420	0.506	0.613	-0.610	1.035
Rndrng_Prvdr_Type_simplified_Obstetrics & Gynecology					
0.8845	0.454	1.949	0.051	-0.005	1.774
Rndrng_Prvdr_Type_simplified_Occupational Therapist in Private Practice					
-1.3569	1.085	-1.250	0.211	-3.484	0.771
Rndrng_Prvdr_Type_simplified_Ophthalmology					
-0.7786	0.577	-1.350	0.177	-1.909	0.352
Rndrng_Prvdr_Type_simplified_Optometry					
-1.7136	0.707	-2.425	0.015	-3.099	-0.329
Rndrng_Prvdr_Type_simplified_Orthopedic Surgery					
-0.2645	0.497	-0.532	0.595	-1.239	0.710
Rndrng_Prvdr_Type_simplified_Other					
0.5608	0.416	1.349	0.177	-0.254	1.376
Rndrng_Prvdr_Type_simplified_Otolaryngology					
0.1444	0.543	0.266	0.790	-0.920	1.209
Rndrng_Prvdr_Type_simplified_Pathology					
-1.2742	0.819	-1.557	0.120	-2.878	0.330
Rndrng_Prvdr_Type_simplified_Physical Medicine and Rehabilitation					
1.3806	0.462	2.986	0.003	0.474	2.287
Rndrng_Prvdr_Type_simplified_Physical Therapist in Private Practice					
-0.4926	0.461	-1.069	0.285	-1.395	0.410

Rndrng_Prvdr_Type_simplified_Physician Assistant					
-0.6620	0.456	-1.450	0.147	-1.557	0.233
Rndrng_Prvdr_Type_simplified_Podiatry					
0.8173	0.452	1.807	0.071	-0.069	1.704
Rndrng_Prvdr_Type_simplified_Psychiatry					
1.3622	0.441	3.087	0.002	0.497	2.227
Rndrng_Prvdr_Type_simplified_Psychologist, Clinical					
-0.1819	0.541	-0.336	0.737	-1.243	0.879
Rndrng_Prvdr_Type_simplified_Pulmonary Disease					
-1.2712	0.819	-1.553	0.120	-2.876	0.333
Rndrng_Prvdr_Type_simplified_Urology					
-0.4419	0.650	-0.680	0.497	-1.716	0.832
Rndrng_Prvdr_State_simplified_AR					
-0.9486	0.497	-1.910	0.056	-1.922	0.025
Rndrng_Prvdr_State_simplified_AZ					
-0.8966	0.373	-2.401	0.016	-1.628	-0.165
Rndrng_Prvdr_State_simplified_CA					
-0.4254	0.271	-1.570	0.116	-0.956	0.106
Rndrng_Prvdr_State_simplified_CO					
-1.9966	0.565	-3.534	0.000	-3.104	-0.889
Rndrng_Prvdr_State_simplified_CT					
-1.2161	0.442	-2.750	0.006	-2.083	-0.349
Rndrng_Prvdr_State_simplified_FL					
-0.3421	0.277	-1.236	0.217	-0.885	0.200
Rndrng_Prvdr_State_simplified_GA					
-0.9316	0.344	-2.707	0.007	-1.606	-0.257
Rndrng_Prvdr_State_simplified_IA					
-1.1106	0.492	-2.258	0.024	-2.075	-0.147
Rndrng_Prvdr_State_simplified_ID					
-1.2467	0.769	-1.621	0.105	-2.754	0.260
Rndrng_Prvdr_State_simplified_IL					
-0.7329	0.310	-2.364	0.018	-1.341	-0.125
Rndrng_Prvdr_State_simplified_IN					
-2.4720	0.633	-3.903	0.000	-3.713	-1.231
Rndrng_Prvdr_State_simplified_KS					
-1.7287	0.640	-2.703	0.007	-2.982	-0.475
Rndrng_Prvdr_State_simplified_KY					
-0.3963	0.359	-1.104	0.270	-1.100	0.307
Rndrng_Prvdr_State_simplified_LA					
-0.3042	0.381	-0.799	0.424	-1.050	0.442
Rndrng_Prvdr_State_simplified_MA					
-1.4332	0.388	-3.698	0.000	-2.193	-0.674
Rndrng_Prvdr_State_simplified_MD					
-1.1798	0.411	-2.870	0.004	-1.986	-0.374
Rndrng_Prvdr_State_simplified_ME					
-0.1915	0.477	-0.402	0.688	-1.126	0.743
Rndrng_Prvdr_State_simplified_MI					
-0.4832	0.304	-1.591	0.112	-1.078	0.112

Rndrng_Prvdr_State_simplified_MN					
-2.0310	0.564	-3.598	0.000	-3.137	-0.925
Rndrng_Prvdr_State_simplified_MO					
-0.4958	0.347	-1.429	0.153	-1.176	0.184
Rndrng_Prvdr_State_simplified_MS					
-0.1067	0.393	-0.272	0.786	-0.876	0.663
Rndrng_Prvdr_State_simplified_NC					
-1.3120	0.371	-3.541	0.000	-2.038	-0.586
Rndrng_Prvdr_State_simplified_NE					
-1.1256	0.575	-1.958	0.050	-2.252	0.001
Rndrng_Prvdr_State_simplified_NH					
-15.6234	710.685	-0.022	0.982	-1408.541	1377.294
Rndrng_Prvdr_State_simplified_NJ					
-1.2470	0.353	-3.534	0.000	-1.939	-0.555
Rndrng_Prvdr_State_simplified_NM					
0.1483	0.431	0.344	0.731	-0.697	0.994
Rndrng_Prvdr_State_simplified_NV					
-0.3481	0.455	-0.765	0.444	-1.240	0.544
Rndrng_Prvdr_State_simplified_NY					
-0.7345	0.285	-2.580	0.010	-1.293	-0.176
Rndrng_Prvdr_State_simplified_OH					
-0.3735	0.292	-1.279	0.201	-0.946	0.199
Rndrng_Prvdr_State_simplified_OK					
-0.6450	0.417	-1.548	0.122	-1.462	0.172
Rndrng_Prvdr_State_simplified_OR					
-0.2742	0.367	-0.748	0.455	-0.993	0.445
Rndrng_Prvdr_State_simplified_Other					
-0.5573	0.319	-1.748	0.081	-1.182	0.068
Rndrng_Prvdr_State_simplified_PA					
-0.6464	0.294	-2.199	0.028	-1.223	-0.070
Rndrng_Prvdr_State_simplified_PR					
-0.3715	0.500	-0.744	0.457	-1.351	0.608
Rndrng_Prvdr_State_simplified_SC					
-1.5645	0.487	-3.212	0.001	-2.519	-0.610
Rndrng_Prvdr_State_simplified_TN					
-0.0384	0.308	-0.125	0.901	-0.641	0.565
Rndrng_Prvdr_State_simplified_TX					
-0.3451	0.277	-1.248	0.212	-0.887	0.197
Rndrng_Prvdr_State_simplified_UT					
-1.9977	0.761	-2.624	0.009	-3.490	-0.506
Rndrng_Prvdr_State_simplified_VA					
-0.7633	0.338	-2.257	0.024	-1.426	-0.100
Rndrng_Prvdr_State_simplified_WA					
-0.6014	0.339	-1.776	0.076	-1.265	0.062
Rndrng_Prvdr_State_simplified_WI					
-1.3261	0.424	-3.128	0.002	-2.157	-0.495
Rndrng_Prvdr_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:.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()
```

ROC Curve – Provider Exclusion (Logistic Regression)

