	Table Of Content :- • 1- Import libraries • 2- Read the dataset files
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	1- import libraries:- import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns
	<pre>import warnings warnings.filterwarnings("ignore") 2- Read the dataset file:-</pre>
	<pre>## Read data files as dataframes with pandas drugs_df</pre>
In [3]:	<pre>insurance_df</pre>
In [4]:	<pre># change column name from physID to doctorID doctor_df = doctor_df.rename(columns={'physID':'doctorID'}) prescriptions_df = prescriptions_df.rename(columns={'physID':'doctorID'})</pre>
	<pre>print('doctor_df : ',doctor_df.columns) print('prescriptions_df : ',prescriptions_df.columns) doctor_df : Index(['doctorID', 'name', 'address', 'phone'], dtype='object') prescriptions_df : Index(['patientID', 'doctorID', 'NDC', 'qty', 'days', 'refills', 'status'], dtype='object')</pre>
	3- Merge Data Files into one Dataframe:- - Merge ALL The DataFrames into one Super DataFrame called Data using Left outer join # 1. Merge patient and prescriptions on 'patientID'
	data = pd.merge(patient_df, prescriptiona_df, on='patientID', how='left') # 2. Merge with doctor information on 'doctorID' data = pd.merge(data, doctor_df, on='doctorID', how='left', suffixes=('_patient', '_doctor')) # 3. Merge with drug information on 'NDC' data = pd.merge(data, drugs_df, on='NDC', how='left', suffixes=('_merged', '_drug')) # 4. Merge with supplier information on 'supID' data = pd.merge(data, supplier_df, on='supID', how='left', suffixes=('_drug', '_supplier')) # 5. Merge with insurance information on 'insurance' (from patient) and 'name' (from insurance) data = pd.merge(data, insurance_df, left_on='insurance', right_on='name', how='left', suffixes=('_merged', '_insurance')) # Display the shape and size of the merged DataFrame print(f"\nShape of the final merged DataFrame: (data.shape)") Shape of the final merged DataFrame: (23, 30) - Show a brief of the dataframe:-
In [44]:	# print top 5 Rows data.head() firstName lastName birthdate address_patient phone_patient gender insurance patientID doctorID NDC expDate supID purchasePrice sellPrice name_supplier address phone_merged name phone_insurance
	0 James Smith 01/01/1987 652 Jill Dr. (868)456-9876 M Molina 1 9.0 23567.0 09/22 1.0 11.23 12.55 Cardinal Health Place, Dublin, OH 43017
	7000 Cardinal Place, Dublin, OH 43017
	2 Huda Saleh 09/22/1999 347 Moss Rd. (313)459-9226 F Alliance 2 2.0 78965.0 05/23 1.0 5.45 6.78 Cardinal Health Place, Dublin, OH 43017
	3 Huda Saleh 09/22/1999 347 Moss Rd. (313)459-9226 F Alliance 2 2.0 23567.0 09/22 1.0 11.23 12.55 Cardinal Health Place, Dublin, OH 43017 6555
	4 Huda Saleh 09/22/1999 347 Moss Rd. (313)459-9226 F Alliance 2 2.0 43234.0 12/22 2.0 33.43 40.33 McKesson Hwy, Irving, TX 75039
In [51]:	
- [TTC[OT]:	18 Fatema Almo 08/06/2004 768 Castle Cir. (313)712-0908 F Molina 13 NaN PriorityHealth (800
	OH 43017 Avery Brandon 02/14/1955 569 Forrest Ln. (134)786-6654 NaN PriorityHealth 14 5.0 45652.0 04/21 2.0 2.34 4.33 McKesson Hwy, Irving, (734)427-2000 PriorityHealth (800)
	TX 75039 21 Jose Martinez 01/19/1988 555 Morris Rd. (976)821-0090 M NaN 15 NaN NaN NaN NaN NaN NaN NaN NaN NaN Na
	5 rows × 30 columns 4- Data Exploration:- print ('Size of data :', data.size)
	<pre>print('Shape of data :',data.shape) print('Dimesnsion of data :',data.ndim) Size of data : 690 Shape of data : (23, 30) Dimesnsion of data : 2</pre>
	3 address_patient 23 non-null object 4 phone_patient 23 non-null object 5 gender 19 non-null object 6 insurance 20 non-null int04 8 doctorI 10 non-null float64 9 NDC 15 non-null float64 10 days 15 non-null float64 11 days 15 non-null float64 12 refills 15 non-null float64 13 status 11 non-null object 13 status 11 non-null object 14 name drug 15 non-null object 15 address_doctor 15 non-null object 16 phone_doctor 15 non-null object 17 brandRame 15 non null object 18 genericName 15 non-null object 19 docage 15 non-null object 20 explate 15 non-null object 21 supID 10 non-null object 22 purchasePrice 15 non-null float64 23 sallPrice 15 non-null object 24 name_supplier 15 non-null object 25 address 15 non-null object 26 phone_macped 15 non-null object 27 name 20 non-null object 28 phone_insurance 20 non-null object 29 prochasePrice 15 non-null object 20 docage 15 non-null object 20 non-null object 21 supID 10 non-null object 22 purchasePrice 15 non-null object 23 sallPrice 15 non-null object 24 name_supplier 15 non-null object 25 address 15 non-null object 26 phone_macped 15 non-null object 27 name 20 non-null object 28 phone_insurance 20 non-null object 29 colar 20 non-null object 30 colar 30 non-null object 40 colar 30 non-null object 40 non-null object 51 support 00 non-null object 61 non-null object 61 non-null object 62 phone_macropulser 5.6 kW 55
<pre>In [75]: Out[75]: In [77]:</pre>	*** Total colorate Control Contr
	67876.0 2 78965.0 1 43234.0 1 34321.0 1 23467.0 1 23467.0 1 23467.0 1 23467.0 1 23467.0 1 2345.0 1 45698.0 1 45698.0 1 17863.0 1 17863.0 1 17865.0 1 Name: NDC, dtype: int64
In [79]:	<pre>data['refills'].value_counts() 0.0 6 5.0 6 3.0 2 2.0 1</pre>
In [49]:	<pre>Name: refills, dtype: int64 # Check Duplicates data.duplicated().sum()</pre>
In [47]:	<pre># Check Missing Values data.isna().sum() firstName 0 lastName 0</pre>
	address_patient 0 0 0 0 0 0 0 0 0 0
	sellPrice 8 name_supplier 8 address 8 phone_merged 8 name 3 phone_insurance 5 coPay 3
	<pre>coPay 3 dtype: int64 daictionary = { # numerical data</pre>
	# Use The Data Dictionary to fill missing Values And Save The Resualts data.fillna(value= daictionary, inplace=True) # Delete missing values
	data.dropna(inplace=True)