

D206__medical__data

September 19, 2021

Research question and description

Data Cleaning Plan

```
[1]: #import necessary packages

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
%matplotlib inline
```

```
[2]: #import raw data as dataframe
df = pd.read_csv('medical_raw_data.csv')
#inspect structure of data
print(df.shape)
```

(10000, 53)

```
[3]: print(df.describe())
```

	Unnamed: 0	CaseOrder	Zip	Lat	Lng \
count	10000.00000	10000.00000	10000.000000	10000.000000	10000.000000
mean	5000.50000	5000.50000	50159.323900	38.751099	-91.243080
std	2886.89568	2886.89568	27469.588208	5.403085	15.205998
min	1.00000	1.00000	610.000000	17.967190	-174.209690
25%	2500.75000	2500.75000	27592.000000	35.255120	-97.352982
50%	5000.50000	5000.50000	50207.000000	39.419355	-88.397230
75%	7500.25000	7500.25000	72411.750000	42.044175	-80.438050
max	10000.00000	10000.00000	99929.000000	70.560990	-65.290170

	Population	Children	Age	Income	VitD_levels \
count	10000.000000	7412.000000	7586.000000	7536.000000	10000.000000
mean	9965.253800	2.098219	53.295676	40484.438268	19.412675
std	14824.758614	2.155427	20.659182	28664.861050	6.723277
min	0.000000	0.000000	18.000000	154.080000	9.519012
25%	694.750000	0.000000	35.000000	19450.792500	16.513171

50%	2769.000000	1.000000	53.000000	33942.280000	18.080560
75%	13945.000000	3.000000	71.000000	54075.235000	19.789740
max	122814.000000	10.000000	89.000000	207249.130000	53.019124

	...	TotalCharge	Additional_charges	Item1	Item2 \
count	...	10000.000000	10000.000000	10000.000000	10000.000000
mean	...	5891.538261	12934.528586	3.518800	3.506700
std	...	3377.558136	6542.601544	1.031966	1.034825
min	...	1256.751699	3125.702716	1.000000	1.000000
25%	...	3253.239465	7986.487642	3.000000	3.000000
50%	...	5852.250564	11573.979365	4.000000	3.000000
75%	...	7614.989701	15626.491033	4.000000	4.000000
max	...	21524.224210	30566.073130	8.000000	7.000000

		Item3	Item4	Item5	Item6	Item7 \
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean		3.511100	3.515100	3.496900	3.522500	3.494000
std		1.032755	1.036282	1.030192	1.032376	1.021405
min		1.000000	1.000000	1.000000	1.000000	1.000000
25%		3.000000	3.000000	3.000000	3.000000	3.000000
50%		4.000000	4.000000	3.000000	4.000000	3.000000
75%		4.000000	4.000000	4.000000	4.000000	4.000000
max		8.000000	7.000000	7.000000	7.000000	7.000000

	Item8
count	10000.000000
mean	3.509700
std	1.042312
min	1.000000
25%	3.000000
50%	3.000000
75%	4.000000
max	7.000000

[8 rows x 26 columns]

```
[4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 53 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Unnamed: 0      10000 non-null  int64
1   CaseOrder       10000 non-null  int64
2   Customer_id     10000 non-null  object
3   Interaction     10000 non-null  object
4   UID             10000 non-null  object
```

5	City	10000	non-null	object
6	State	10000	non-null	object
7	County	10000	non-null	object
8	Zip	10000	non-null	int64
9	Lat	10000	non-null	float64
10	Lng	10000	non-null	float64
11	Population	10000	non-null	int64
12	Area	10000	non-null	object
13	Timezone	10000	non-null	object
14	Job	10000	non-null	object
15	Children	7412	non-null	float64
16	Age	7586	non-null	float64
17	Education	10000	non-null	object
18	Employment	10000	non-null	object
19	Income	7536	non-null	float64
20	Marital	10000	non-null	object
21	Gender	10000	non-null	object
22	ReAdmis	10000	non-null	object
23	VitD_levels	10000	non-null	float64
24	Doc_visits	10000	non-null	int64
25	Full_meals_eaten	10000	non-null	int64
26	VitD_supp	10000	non-null	int64
27	Soft_drink	7533	non-null	object
28	Initial_admin	10000	non-null	object
29	HighBlood	10000	non-null	object
30	Stroke	10000	non-null	object
31	Complication_risk	10000	non-null	object
32	Overweight	9018	non-null	float64
33	Arthritis	10000	non-null	object
34	Diabetes	10000	non-null	object
35	Hyperlipidemia	10000	non-null	object
36	BackPain	10000	non-null	object
37	Anxiety	9016	non-null	float64
38	Allergic_rhinitis	10000	non-null	object
39	Reflux_esophagitis	10000	non-null	object
40	Asthma	10000	non-null	object
41	Services	10000	non-null	object
42	Initial_days	8944	non-null	float64
43	TotalCharge	10000	non-null	float64
44	Additional_charges	10000	non-null	float64
45	Item1	10000	non-null	int64
46	Item2	10000	non-null	int64
47	Item3	10000	non-null	int64
48	Item4	10000	non-null	int64
49	Item5	10000	non-null	int64
50	Item6	10000	non-null	int64
51	Item7	10000	non-null	int64
52	Item8	10000	non-null	int64

```
dtypes: float64(11), int64(15), object(27)
memory usage: 4.0+ MB
```

```
[5]: df.head()
```

```
[5]:   Unnamed: 0  CaseOrder Customer_id Interaction \
0           1           1      C412403  8cd49b13-f45a-4b47-a2bd-173ffa932c2f
1           2           2      Z919181  d2450b70-0337-4406-bdbb-bc1037f1734c
2           3           3      F995323  a2057123-abf5-4a2c-abad-8ffe33512562
3           4           4      A879973  1dec528d-eb34-4079-adce-0d7a40e82205
4           5           5      C544523  5885f56b-d6da-43a3-8760-83583af94266

      UID      City State      County  Zip \
0  3a83ddb66e2ae73798bdf1d705dc0932      Eva    AL      Morgan  35621
1  176354c5eef714957d486009feabf195  Marianna  FL      Jackson  32446
2  e19a0fa00aeda885b8a436757e889bc9  Sioux Falls  SD      Minnehaha  57110
3  cd17d7b6d152cb6f23957346d11c3f07  New Richland  MN      Waseca  56072
4  d2f0425877b10ed6bb381f3e2579424a  West Point  VA  King William  23181

      Lat  ...  TotalCharge  Additional_charges  Item1  Item2  Item3  Item4  \
0  34.34960  ...  3191.048774      17939.403420      3      3      2      2
1  30.84513  ...  4214.905346      17612.998120      3      4      3      4
2  43.54321  ...  2177.586768      17505.192460      2      4      4      4
3  43.89744  ...  2465.118965      12993.437350      3      5      5      3
4  37.59894  ...  1885.655137      3716.525786      2      1      3      3

      Item5  Item6  Item7  Item8
0         4      3      3      4
1         4      4      3      3
2         3      4      3      3
3         4      5      5      5
4         5      3      4      3
```

```
[5 rows x 53 columns]
```

- remove redundant columns, and columns that will not contribute meaningfully to analysis
- column Unnamed: 0 is removed because it is functionally identical to CaseOrder, Lat and Lng are
- dropped due to being personally identifiable information about patient that does not contribute
- more meaningfully to analysis than anonymized columns such as Area and Timezone already do.
- other columns are dropped due to being internal system labels that are not useful for analysis.
- set index to column CaseOrder

```
[6]: df = df.drop(columns=["Unnamed: 0", "Customer_id", "Interaction", "UID", "Lat", "Lng"])
      ↪ "Lng"])
```

```
[7]: df.rename(columns={"CaseOrder" : "Case_order"}, inplace=True)
df = df.set_index("Case_order", drop = True)
df.head()
```

```
[7]:
```

	City	State	County	Zip	Population	Area	\
Case_order							
1	Eva	AL	Morgan	35621	2951	Suburban	
2	Marianna	FL	Jackson	32446	11303	Urban	
3	Sioux Falls	SD	Minnehaha	57110	17125	Suburban	
4	New Richland	MN	Waseca	56072	2162	Suburban	
5	West Point	VA	King William	23181	5287	Rural	

	Timezone	Job	Children	\
Case_order				
1	America/Chicago	Psychologist, sport and exercise	1.0	
2	America/Chicago	Community development worker	3.0	
3	America/Chicago	Chief Executive Officer	3.0	
4	America/Chicago	Early years teacher	0.0	
5	America/New_York	Health promotion specialist	NaN	

	Age	...	TotalCharge	Additional_charges	Item1	Item2	Item3	\
Case_order		...						
1	53.0	...	3191.048774	17939.403420	3	3	2	
2	51.0	...	4214.905346	17612.998120	3	4	3	
3	53.0	...	2177.586768	17505.192460	2	4	4	
4	78.0	...	2465.118965	12993.437350	3	5	5	
5	22.0	...	1885.655137	3716.525786	2	1	3	

	Item4	Item5	Item6	Item7	Item8
Case_order					
1	2	4	3	3	4
2	4	4	4	3	3
3	4	3	4	3	3
4	3	4	5	5	5
5	3	5	3	4	3

[5 rows x 46 columns]

```
[8]: df.columns
```

```
[8]: Index(['City', 'State', 'County', 'Zip', 'Population', 'Area', 'Timezone',
        'Job', 'Children', 'Age', 'Education', 'Employment', 'Income',
        'Marital', 'Gender', 'ReAdmis', 'VitD_levels', 'Doc_visits',
        'Full_meals_eaten', 'VitD_supp', 'Soft_drink', 'Initial_admin',
        'HighBlood', 'Stroke', 'Complication_risk', 'Overweight', 'Arthritis',
        'Diabetes', 'Hyperlipidemia', 'BackPain', 'Anxiety',
        'Allergic_rhinitis', 'Reflux_esophagitis', 'Asthma', 'Services',
        'Initial_days', 'TotalCharge', 'Additional_charges', 'Item1', 'Item2',
        'Item3', 'Item4', 'Item5', 'Item6', 'Item7', 'Item8'],
        dtype='object')
```

```
[9]: #standardise column names, and update column names to be more descriptive
df.rename(columns={"Marital": "Mariage_status", "ReAdmis": "Readmited",
↳ "VitD_supp": "VitD_supplements",
↳ "Soft_drink": "Habitual_soft_drink_use", "BackPain":
↳ "Back_pain", "Services":
↳ "Primary_service_recived", "HighBlood":
↳ "High_blood_pressure", "TotalCharge": "Total_charge",
↳ "Item1": "Survey_timely_admission", "Item2":
↳ "Survey_timely_treatment",
↳ "Item3": "Survey_timely_visits", "Item4":
↳ "Survey_reliability",
↳ "Item5": "Survey_options", "Item6": "Survey_hours",
↳ "Item7": "Survey_courtesy", "Item8":
↳ "Survey_active_listening"}, inplace=True)
```

```
[10]: #check for duplicated rows
df.duplicated().any()
```

[10]: False

```
[11]: #check if any rows contain only null values
df.isnull().all(axis=1).any()
```

[11]: False

```
[12]: #determine which columns contain null values
contains_missing = df.loc[:,df.isnull().any()].copy()
contains_missing
```

```
[12]:
```

	Children	Age	Income	Habitual_soft_drink_use	Overweight	\
Case_order						
1	1.0	53.0	86575.93	NaN	0.0	
2	3.0	51.0	46805.99	No	1.0	
3	3.0	53.0	14370.14	No	1.0	
4	0.0	78.0	39741.49	No	0.0	
5	NaN	22.0	1209.56	Yes	0.0	
...	
9996	NaN	25.0	45967.61	No	NaN	
9997	4.0	87.0	14983.02	No	1.0	
9998	3.0	NaN	65917.81	Yes	1.0	
9999	3.0	43.0	29702.32	No	1.0	
10000	8.0	NaN	62682.63	No	1.0	

	Anxiety	Initial_days
Case_order		
1	1.0	10.585770
2	NaN	15.129562

3	NaN	4.772177
4	NaN	1.714879
5	0.0	1.254807
...
9996	1.0	51.561217
9997	0.0	68.668237
9998	1.0	NaN
9999	0.0	63.356903
10000	0.0	70.850592

[10000 rows x 7 columns]

```
[13]: #use mode to impute missing values in catagorical variables(columns:
      ↳ Habitual_soft_drink_use, Overweight, Anxiety),
      #mode is used due to the variables being catagorical
      contains_missing['Habitual_soft_drink_use'].
      ↳ fillna(contains_missing['Habitual_soft_drink_use'].mode()[0], inplace = True)
      contains_missing['Overweight'].fillna(contains_missing['Overweight'].mode()[0],
      ↳ inplace = True)
      contains_missing['Anxiety'].fillna(contains_missing['Anxiety'].mode()[0],
      ↳ inplace = True)
      contains_missing
```

```
[13]:
```

	Children	Age	Income	Habitual_soft_drink_use	Overweight	\
Case_order						
1	1.0	53.0	86575.93	No	0.0	
2	3.0	51.0	46805.99	No	1.0	
3	3.0	53.0	14370.14	No	1.0	
4	0.0	78.0	39741.49	No	0.0	
5	NaN	22.0	1209.56	Yes	0.0	
...	
9996	NaN	25.0	45967.61	No	1.0	
9997	4.0	87.0	14983.02	No	1.0	
9998	3.0	NaN	65917.81	Yes	1.0	
9999	3.0	43.0	29702.32	No	1.0	
10000	8.0	NaN	62682.63	No	1.0	

	Anxiety	Initial_days
Case_order		
1	1.0	10.585770
2	0.0	15.129562
3	0.0	4.772177
4	0.0	1.714879
5	0.0	1.254807
...
9996	1.0	51.561217
9997	0.0	68.668237

```
9998      1.0      NaN
9999      0.0    63.356903
10000     0.0    70.850592
```

```
[10000 rows x 7 columns]
```

```
[14]: contains_missing['Habitual_soft_drink_use'].value_counts()
```

```
[14]: No      8056
      Yes     1944
      Name: Habitual_soft_drink_use, dtype: int64
```

```
[15]: contains_missing['Overweight'].value_counts()
```

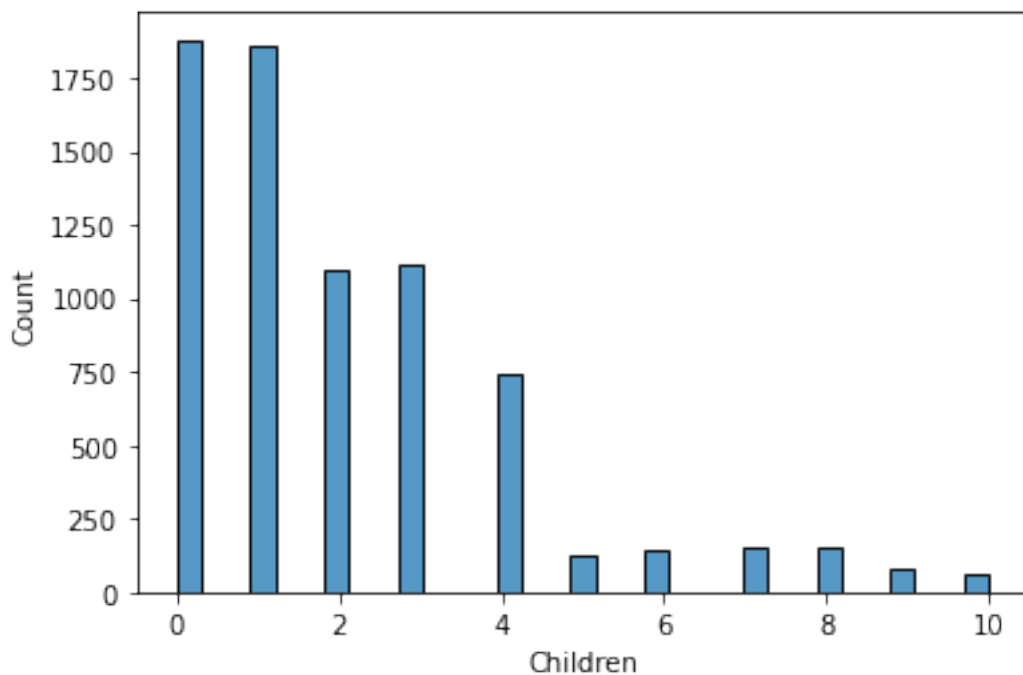
```
[15]: 1.0     7377
      0.0     2623
      Name: Overweight, dtype: int64
```

```
[16]: contains_missing['Anxiety'].value_counts()
```

```
[16]: 0.0     7094
      1.0     2906
      Name: Anxiety, dtype: int64
```

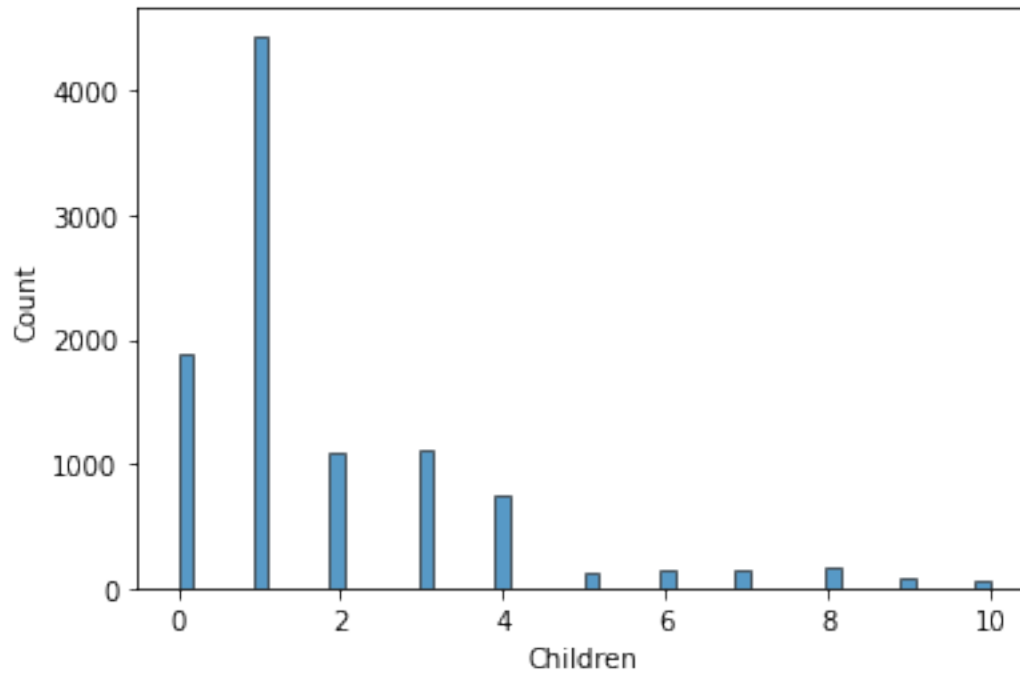
```
[17]: sns.histplot(contains_missing['Children'])
```

```
[17]: <AxesSubplot:xlabel='Children', ylabel='Count'>
```



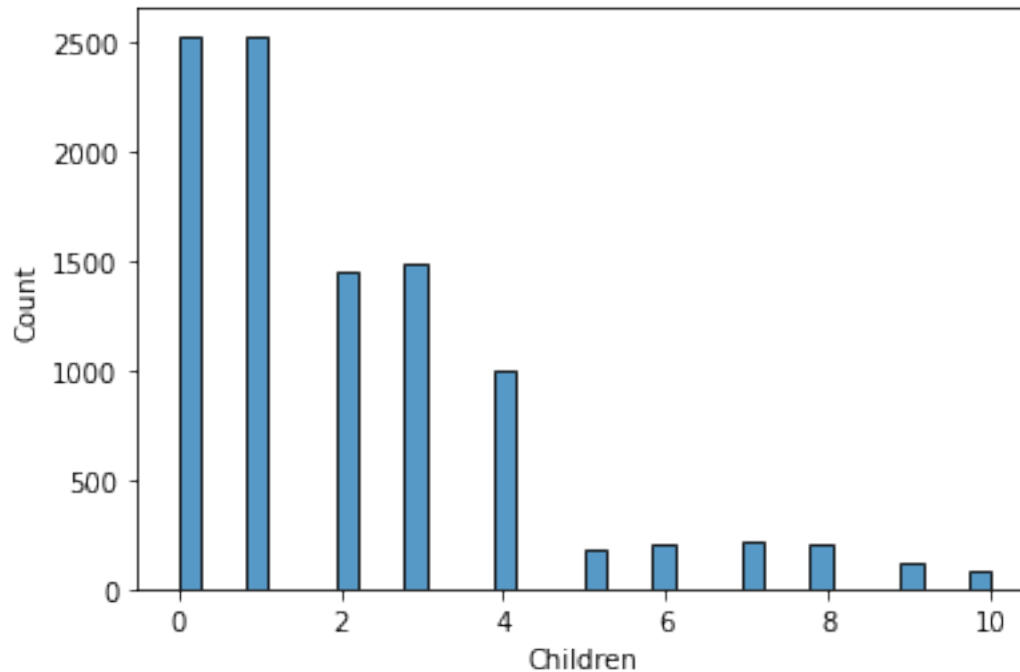

```
[18]: sns.histplot(contains_missing['Children'].fillna(contains_missing['Children'].  
↳median()))
```

```
[18]: <AxesSubplot:xlabel='Children', ylabel='Count'>
```



```
[19]: sns.histplot(contains_missing['Children'].interpolate(method='pad'))
```

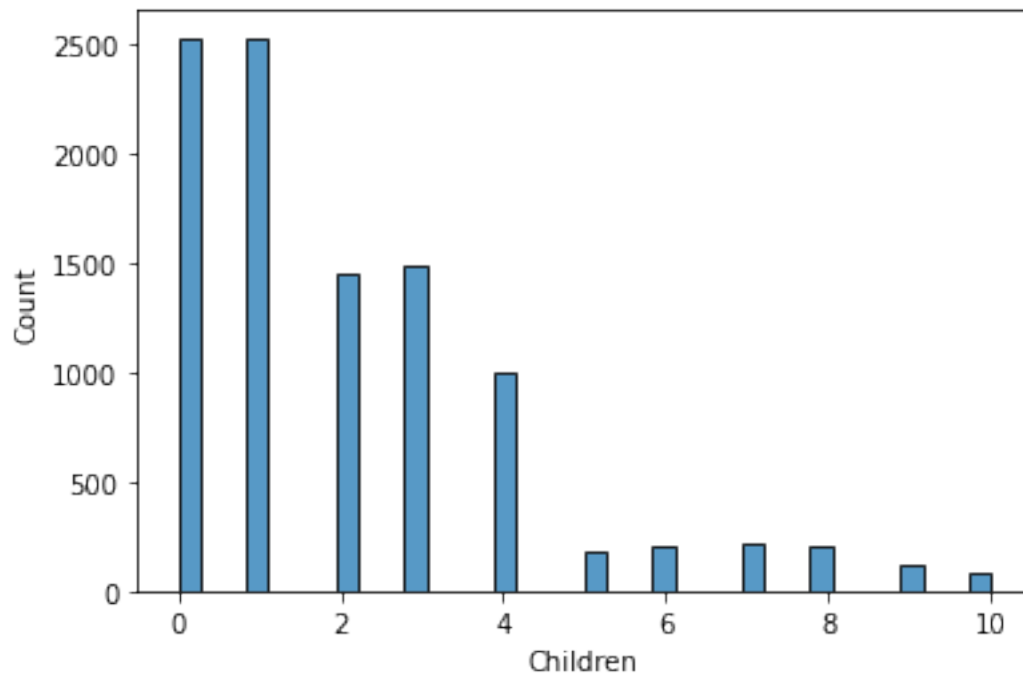
```
[19]: <AxesSubplot:xlabel='Children', ylabel='Count'>
```



```
[20]: #use interpolation to impune missing data for number of children. pad method is used to avoid adding values that are
      #not whole numbers, interpolation is used because data skews to the right and
      #the amount of data points equaling 1 to more than double
      contains_missing['Children'].interpolate(method='pad', inplace=True)
```

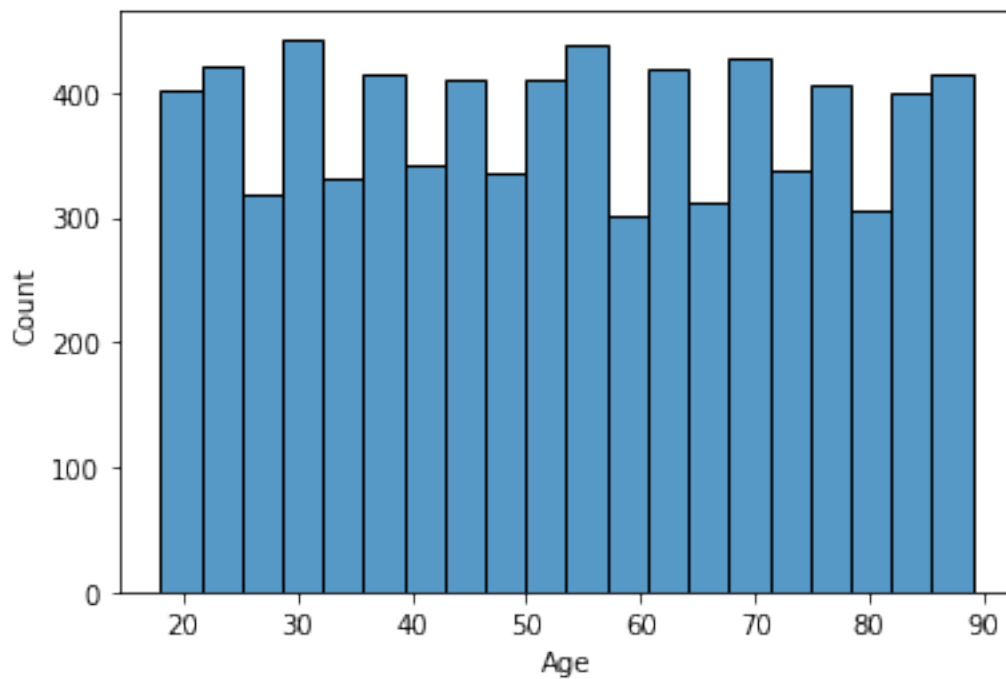
```
[21]: sns.histplot(contains_missing['Children'])
```

```
[21]: <AxesSubplot:xlabel='Children', ylabel='Count'>
```



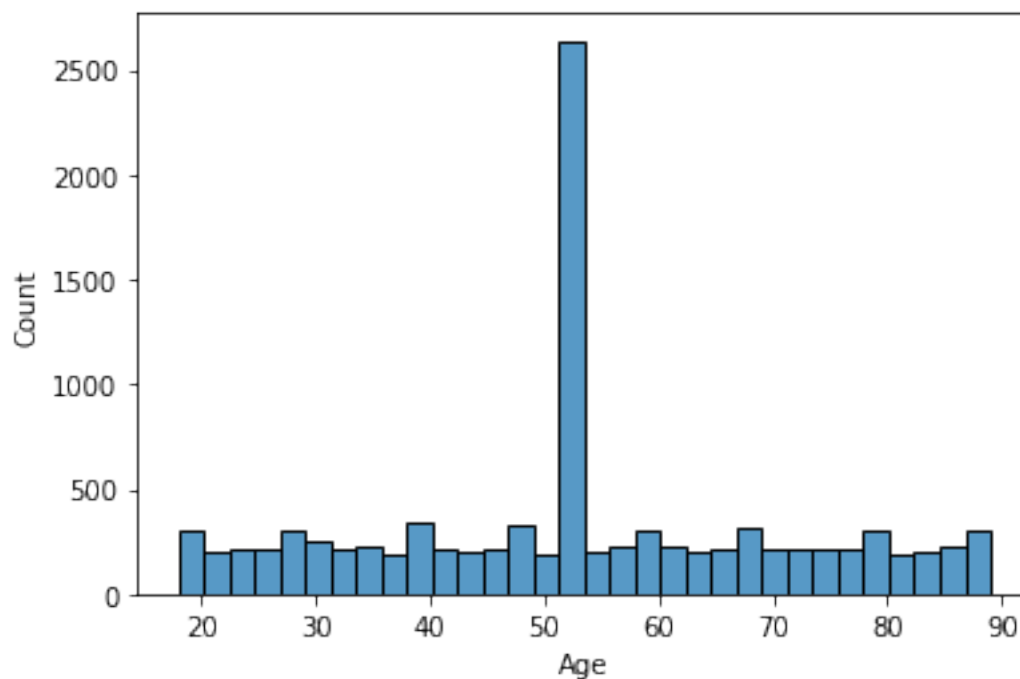
```
[22]: sns.histplot(contains_missing['Age'])
```

```
[22]: <AxesSubplot:xlabel='Age', ylabel='Count'>
```



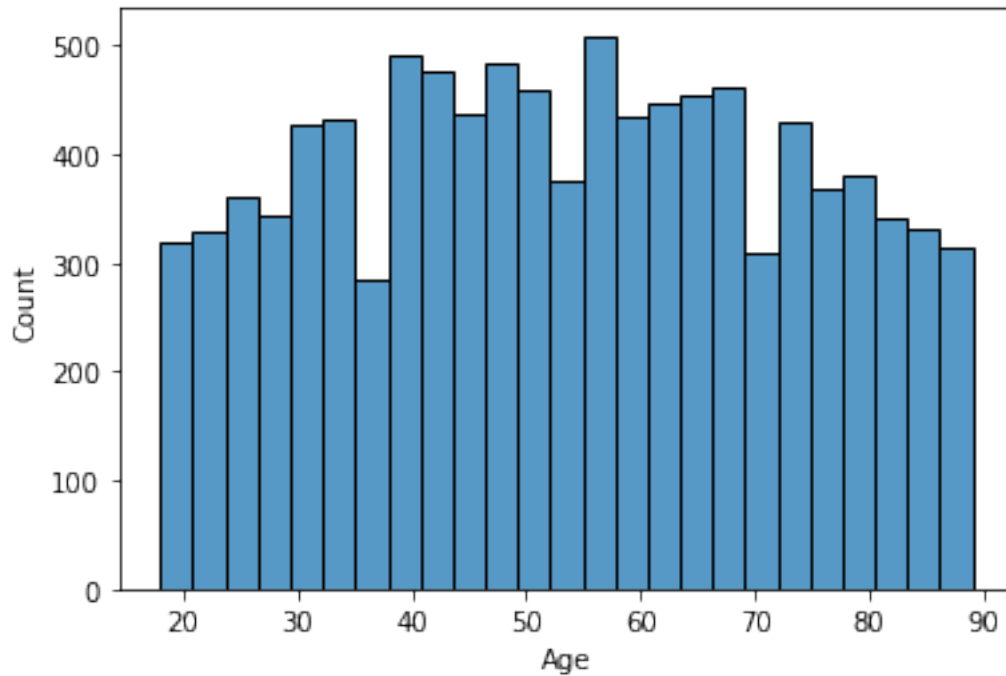
```
[23]: sns.histplot(contains_missing['Age'].fillna(contains_missing['Age'].mean()))
```

```
[23]: <AxesSubplot:xlabel='Age', ylabel='Count'>
```



```
[24]: sns.histplot(contains_missing['Age'].interpolate())
```

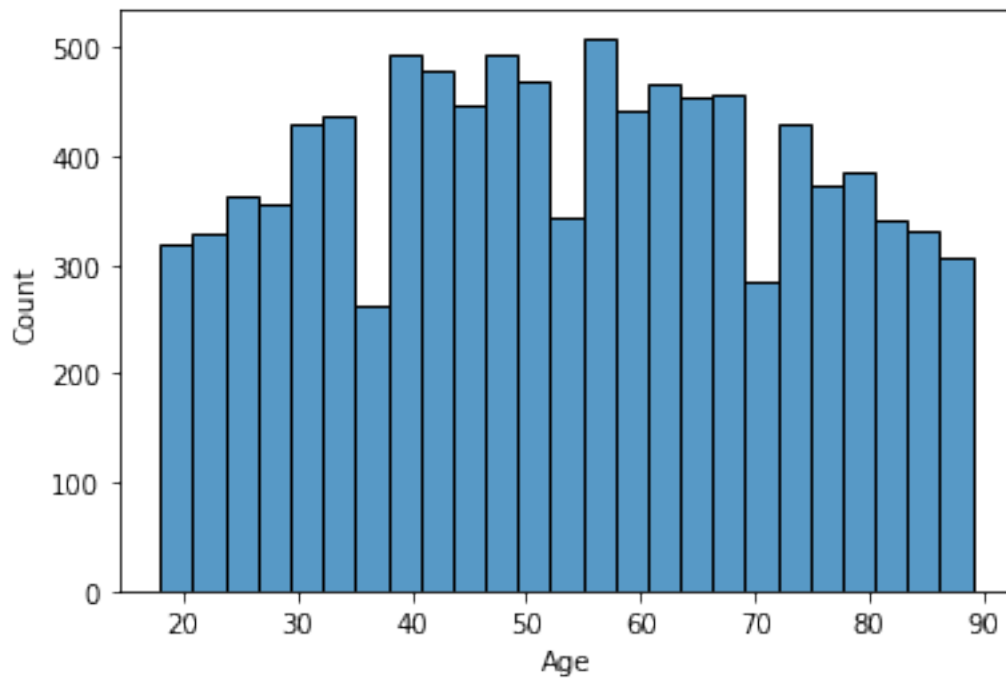
```
[24]: <AxesSubplot:xlabel='Age', ylabel='Count'>
```



```
[25]: #use interpolation to impune missing data for number patient age. interpolation
      ↪ is used because histogram revealed
      #that data is evenly distributed, and using mean created a drastic change in
      ↪ the distribution.
contains_missing.interpolate(inplace=True)
contains_missing['Age'] = contains_missing['Age'].astype(int)
```

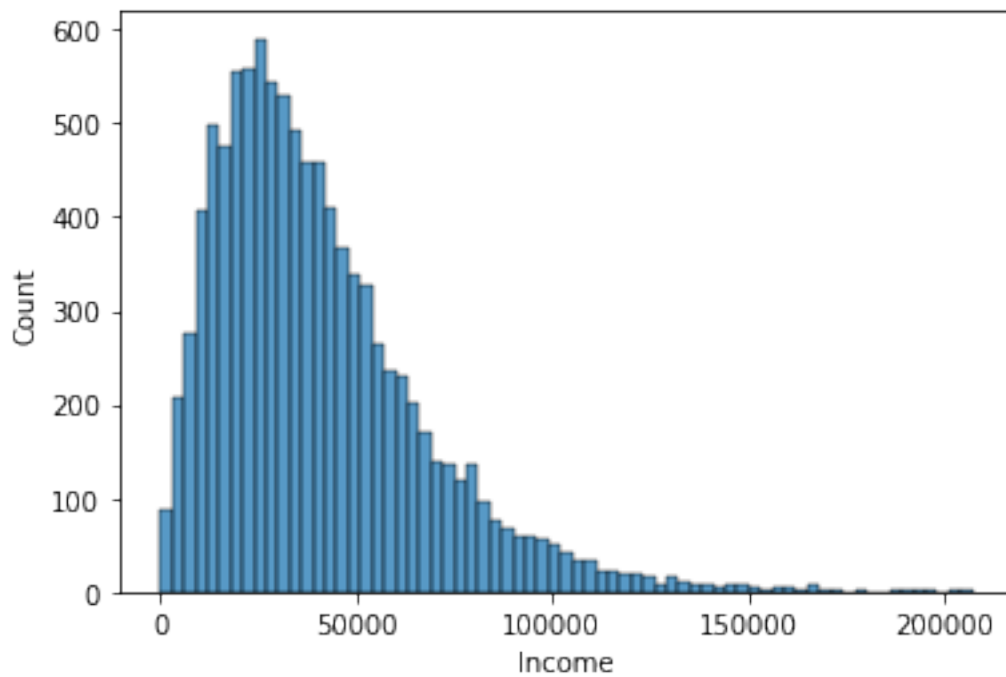
```
[26]: sns.histplot(contains_missing['Age'])
```

```
[26]: <AxesSubplot:xlabel='Age', ylabel='Count'>
```



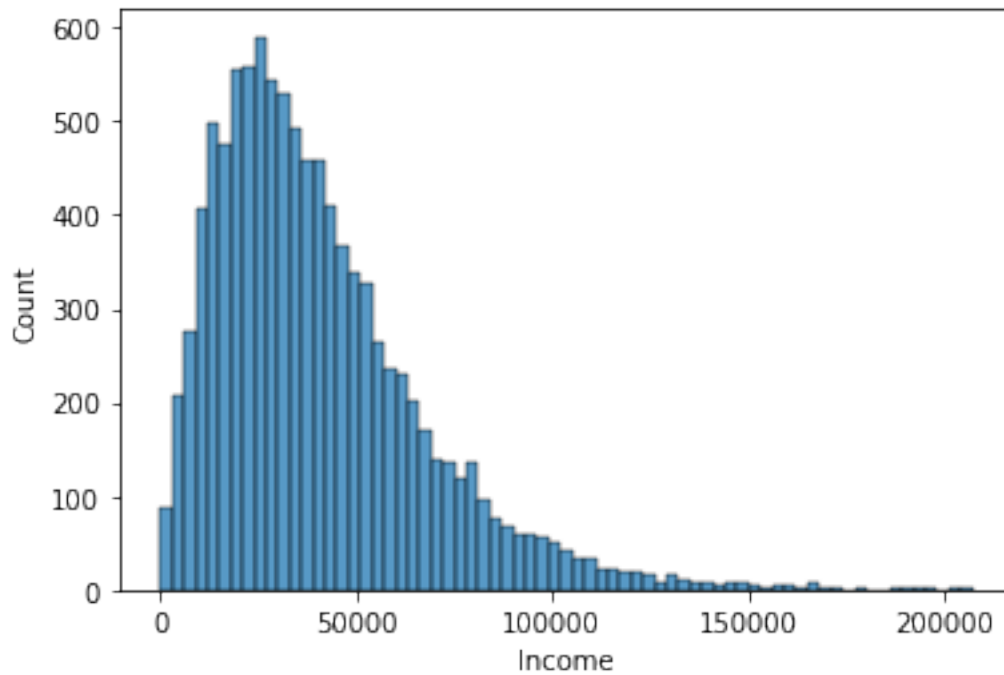
```
[27]: sns.histplot(contains_missing['Income'])
```

```
[27]: <AxesSubplot:xlabel='Income', ylabel='Count'>
```



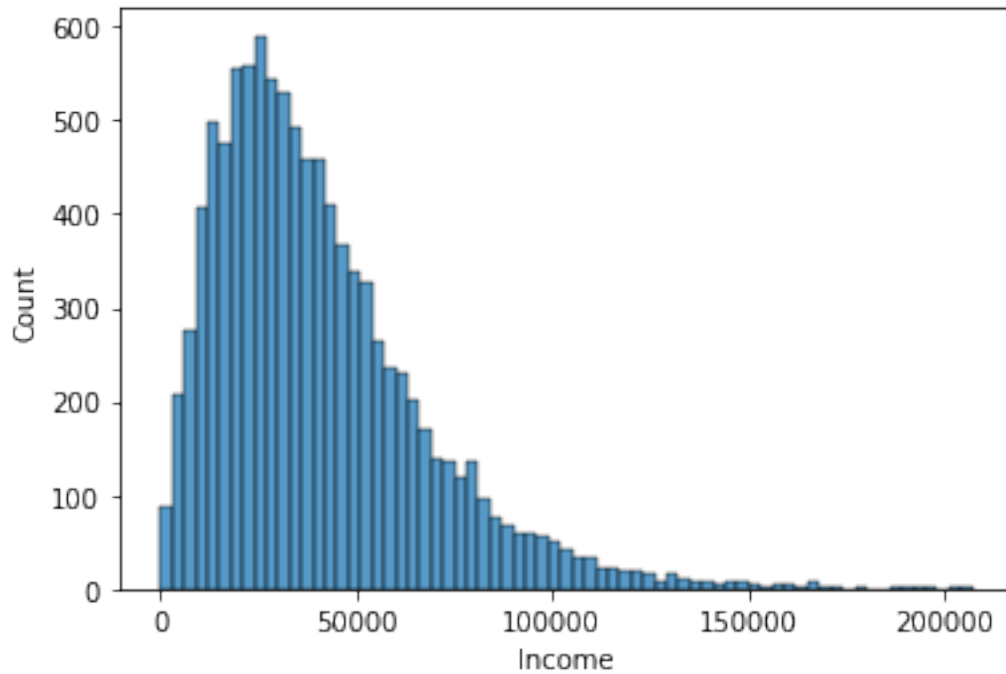
```
[28]: sns.histplot(contains_missing['Income'].fillna(contains_missing['Income'].  
↳median()))
```

```
[28]: <AxesSubplot:xlabel='Income', ylabel='Count'>
```



```
[29]: sns.histplot(contains_missing['Income'].interpolate())
```

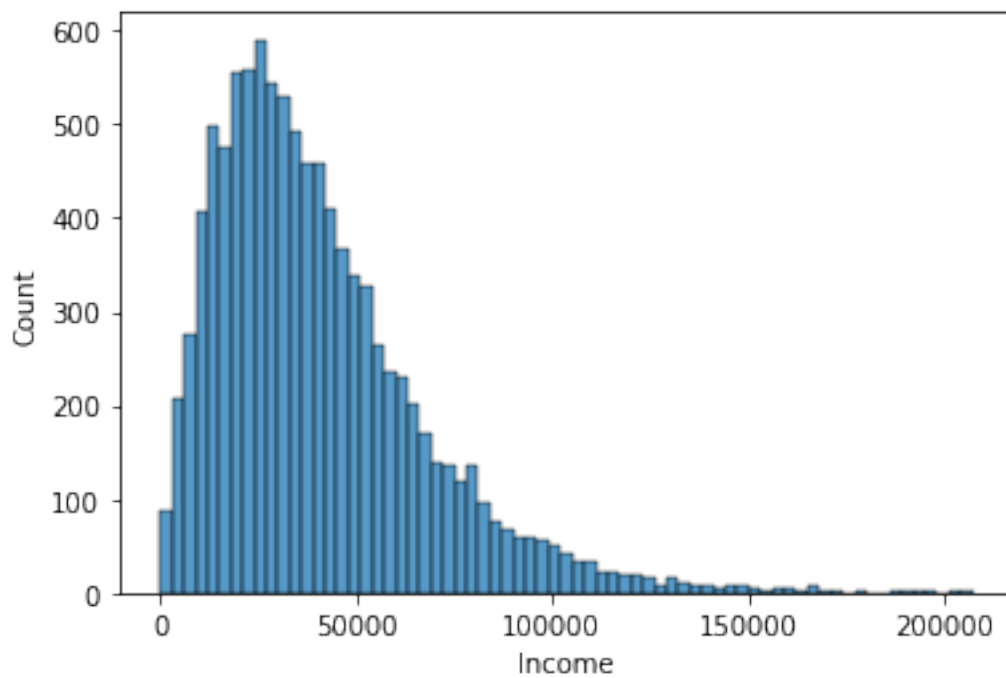
```
[29]: <AxesSubplot:xlabel='Income', ylabel='Count'>
```



```
[30]: #use median value to fill missing values in income column. median was chosen as
      ↳ imputation method because histogram
      #reveled that data skews to the right, and there was no discernible difference
      ↳ bettween imputation and interpolation
      #in maintaining distrobution
contains_missing['Income'].fillna(contains_missing['Income'].median(), inplace_
      ↳ = True)
```

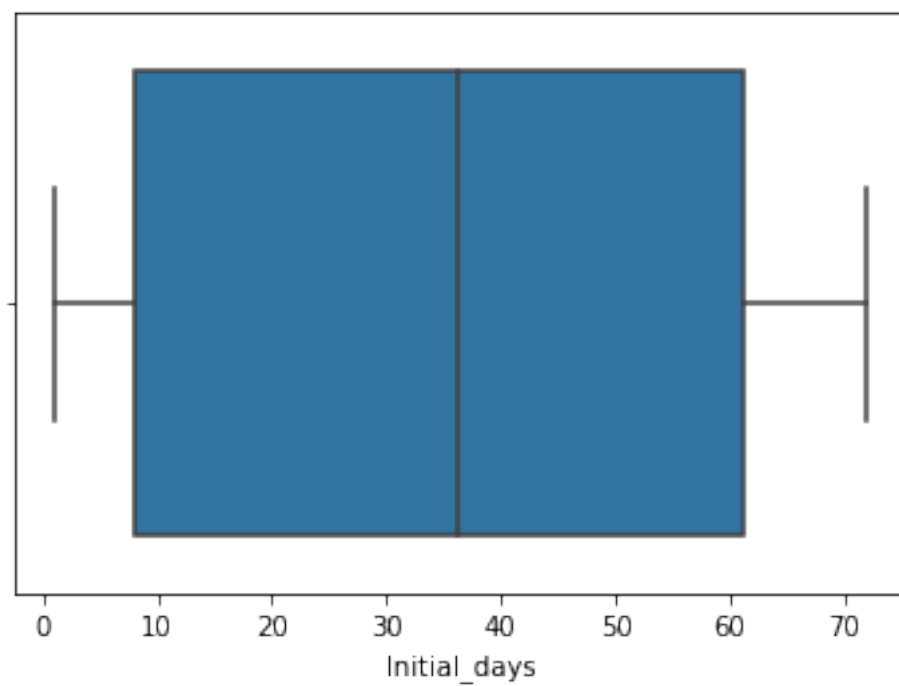
```
[31]: sns.histplot(contains_missing['Income'])
```

```
[31]: <AxesSubplot:xlabel='Income', ylabel='Count'>
```

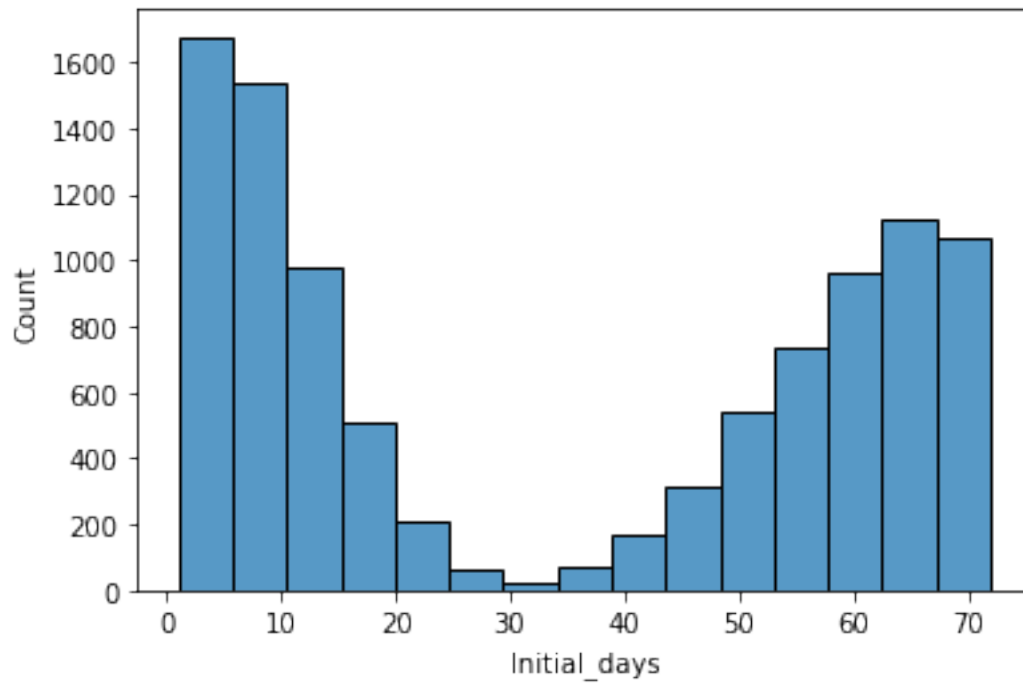
```
[32]: sns.boxplot(x=contains_missing['Initial_days'])
```

```
[32]: <AxesSubplot:xlabel='Initial_days'>
```



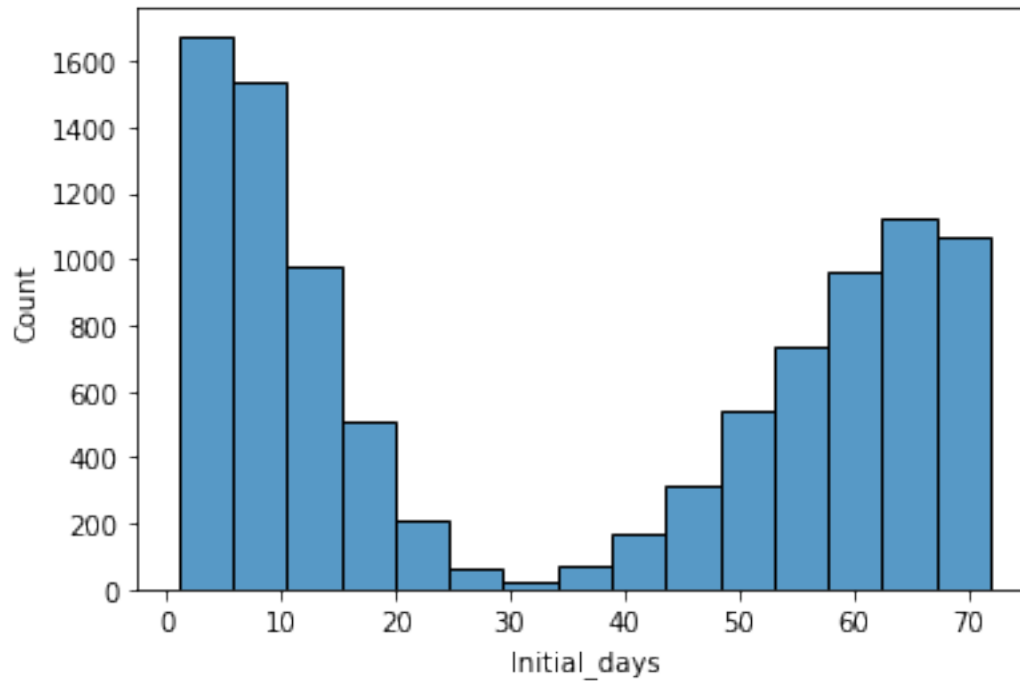
```
[33]: sns.histplot(contains_missing['Initial_days'])
```

```
[33]: <AxesSubplot:xlabel='Initial_days', ylabel='Count'>
```



```
[34]: sns.histplot(contains_missing['Initial_days'],  
→fillna(contains_missing['Initial_days'].mean()))
```

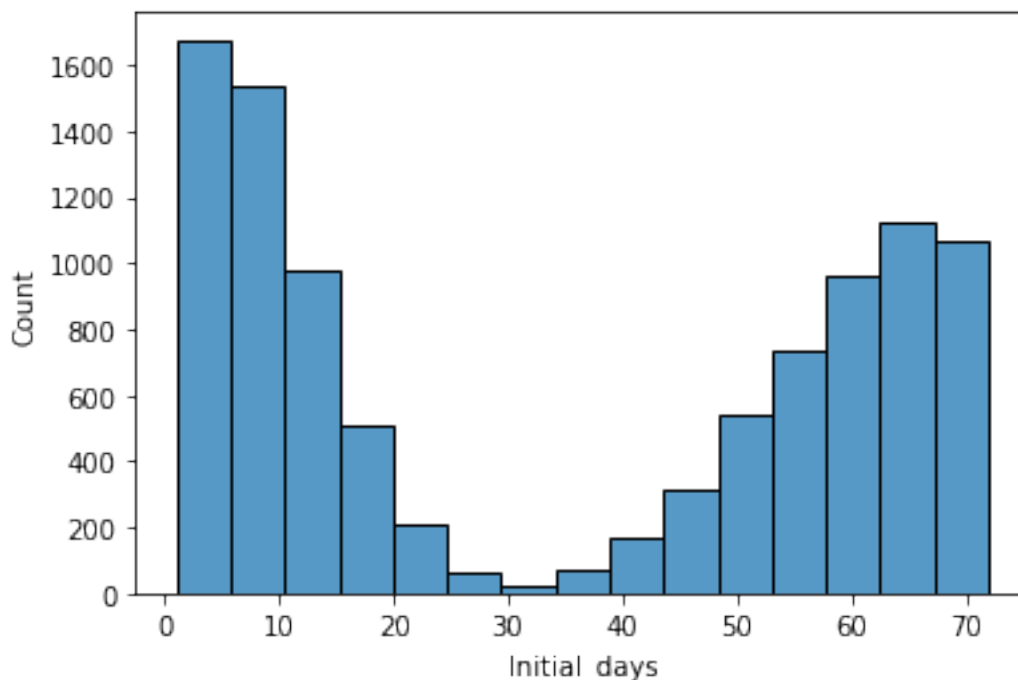
```
[34]: <AxesSubplot:xlabel='Initial_days', ylabel='Count'>
```



```
[35]: #use mean to impune missing data for number patient age. mean is used because
      ↪ histogram and box plot revealed
      #that data has a bimodal distribution, and using mean maintains this
      ↪ distribution.
      contains_missing['Initial_days'].fillna(contains_missing['Initial_days'].
      ↪ mean(), inplace=True)
```

```
[36]: sns.histplot(contains_missing['Initial_days'])
```

```
[36]: <AxesSubplot:xlabel='Initial_days', ylabel='Count'>
```



```
[37]: #replace columns in original dataframe with corrected values in
      ↪ contains_missing dataframe
for x in contains_missing:
    df[x] = contains_missing[x]
```

```
[38]: #display unique values of each column
for col in df:
    print(col + ', ', df[col].dtypes, ' : ')
    print(df[col].unique())
```

```
City, object :
['Eva' 'Marianna' 'Sioux Falls' ... 'Milmay' 'Quinn' 'Coraopolis']
State, object :
['AL' 'FL' 'SD' 'MN' 'VA' 'OK' 'OH' 'MS' 'WI' 'IA' 'CA' 'IN' 'MO' 'MI'
 'NE' 'PA' 'AR' 'WV' 'KS' 'MA' 'KY' 'NY' 'VT' 'DC' 'IL' 'ND' 'SC' 'AK'
 'NM' 'NH' 'GA' 'NC' 'MD' 'TN' 'WA' 'TX' 'CO' 'NJ' 'LA' 'OR' 'AZ' 'ME'
 'ID' 'UT' 'RI' 'MT' 'PR' 'NV' 'CT' 'HI' 'WY' 'DE']
County, object :
['Morgan' 'Jackson' 'Minnehaha' ... 'Navarro' 'Los Alamos' 'Sterling']
Zip, int64 :
[35621 32446 57110 ... 8340 57775 15108]
Population, int64 :
[ 2951 11303 17125 ... 8368 7908 41524]
Area, object :
['Suburban' 'Urban' 'Rural']
```

Timezone, object :

['America/Chicago' 'America/New_York' 'America/Los_Angeles'
'America/Indiana/Indianapolis' 'America/Detroit' 'America/Denver'
'America/Nome' 'America/Anchorage' 'America/Phoenix' 'America/Boise'
'America/Puerto_Rico' 'America/Yakutat' 'Pacific/Honolulu'
'America/Menominee' 'America/Kentucky/Louisville'
'America/Indiana/Vincennes' 'America/Toronto' 'America/Indiana/Marengo'
'America/Indiana/Winamac' 'America/Indiana/Tell_City' 'America/Sitka'
'America/Indiana/Knox' 'America/North_Dakota/New_Salem'
'America/Indiana/Vevay' 'America/Adak' 'America/North_Dakota/Beulah']

Job, object :

['Psychologist, sport and exercise' 'Community development worker'
'Chief Executive Officer' 'Early years teacher'
'Health promotion specialist' 'Corporate treasurer' 'Hydrologist'
'Psychiatric nurse' 'Computer games developer'
'Production assistant, radio' 'Contractor'
'Surveyor, planning and development'
'English as a second language teacher' 'Actuary' 'Media planner'
'Fast food restaurant manager' 'Horticulturist, commercial'
'Secretary, company' 'Designer, graphic' 'Personnel officer'
'Telecommunications researcher' 'Restaurant manager, fast food'
'Surveyor, minerals' 'Architectural technologist'
'Therapist, speech and language' 'Accounting technician'
'Glass blower/designer' 'Travel agency manager' 'Illustrator'
'Police officer' 'Accountant, chartered public finance'
'Sport and exercise psychologist' 'Pensions consultant'
'Community education officer' 'Radio producer'
'Designer, television/film set' 'Conference centre manager'
'Advertising account executive' 'Civil Service fast streamer'
'Training and development officer' 'Buyer, retail' 'Event organiser'
'IT technical support officer'
'Historic buildings inspector/conservation officer'
'Research scientist (physical sciences)' 'Games developer'
'Manufacturing engineer' 'Embryologist, clinical' 'Merchant navy officer'
'Television floor manager' 'Web designer' 'Industrial buyer' 'Aid worker'
'Systems developer' 'Probation officer'
'Scientific laboratory technician' 'Environmental health practitioner'
'Prison officer' 'Naval architect' 'Pilot, airline'
'Medical sales representative' 'Learning disability nurse'
'Agricultural engineer' 'Multimedia programmer' 'Cartographer'
'Company secretary' 'Operations geologist' 'Conservation officer, nature'
'Therapist, art' 'Therapist, sports' 'Oncologist'
'Armed forces logistics/support/administrative officer' 'Podiatrist'
'Translator' 'Geochemist' 'Engineer, technical sales'
'Production designer, theatre/television/film' 'Site engineer'
'Teacher, primary school' 'Clinical molecular geneticist'
'Armed forces operational officer' 'Careers information officer'
'Camera operator' 'Engineer, aeronautical' 'Learning mentor']

'Neurosurgeon' 'Clothing/textile technologist' 'Financial controller'
 'Education officer, museum' 'Set designer'
 'Accountant, chartered certified' 'Solicitor' 'Forensic psychologist'
 'Outdoor activities/education manager' 'Heritage manager'
 'Hospital doctor' 'Engineer, chemical' 'Musician'
 'Engineer, control and instrumentation' 'Engineer, mining'
 'Editor, commissioning' 'Sports development officer' 'Teacher, music'
 'Nurse, children's' 'Editor, film/video' 'Acupuncturist' 'Data scientist'
 'Tax inspector' 'Engineer, maintenance' 'Radiographer, therapeutic'
 'Surveyor, commercial/residential' 'Engineer, civil (contracting)'
 'Therapist, nutritional' 'Public affairs consultant' 'Warehouse manager'
 'Consulting civil engineer' 'Museum/gallery exhibitions officer'
 'Risk manager' 'Air traffic controller' 'Health service manager'
 'Teacher, adult education' 'Theatre stage manager'
 'Designer, fashion/clothing' 'Engineer, site' 'Psychologist, counselling'
 'Product/process development scientist' 'Financial adviser'
 'Quarry manager' 'Librarian, public' 'Presenter, broadcasting'
 'Structural engineer' 'Trade mark attorney' 'Amenity horticulturist'
 'Building services engineer' 'Primary school teacher' 'Network engineer'
 'Psychotherapist, child' 'Archaeologist' 'Publishing rights manager'
 'Economist' 'Herbalist' 'Legal secretary'
 'Engineer, manufacturing systems' 'Psychologist, occupational'
 'Journalist, broadcasting' 'Lexicographer' 'Clinical psychologist'
 'Scientist, water quality'
 'Chartered legal executive (England and Wales)' 'Statistician'
 'Chartered accountant' 'Operational investment banker'
 'Nutritional therapist' 'Actor' 'Ecologist' 'Conservator, furniture'
 'Archivist' 'Industrial/product designer' 'Air broker' 'Sports coach'
 'Chief Technology Officer' 'Arts administrator' 'Restaurant manager'
 'Editorial assistant' 'Cytogeneticist' 'Scientist, marine'
 'Surveyor, quantity' 'Designer, exhibition/display' 'Curator'
 'Human resources officer' 'Osteopath' 'Therapist, music'
 'Volunteer coordinator' 'Office manager' 'Research officer, government'
 'Quality manager' 'Artist' 'Museum education officer'
 'Exercise physiologist'
 'Administrator, charities/voluntary organisations' 'Purchasing manager'
 'Therapeutic radiographer' 'Farm manager' 'Tour manager' 'Writer'
 'Designer, industrial/product' 'Science writer' 'Engineer, biomedical'
 'Development worker, international aid' 'Journalist, newspaper'
 'Multimedia specialist' 'Dealer' 'Water engineer'
 'Scientist, clinical (histocompatibility and immunogenetics)'
 'Special effects artist' 'Engineer, agricultural'
 'Corporate investment banker' 'Best boy'
 'Production assistant, television' 'Chiropractor' 'Jewellery designer'
 'Energy engineer' 'Scientist, forensic' 'Biomedical engineer'
 'Insurance account manager' 'Occupational psychologist'
 'Diagnostic radiographer' 'Banker' 'Medical technical officer'
 'Quantity surveyor' 'Biochemist, clinical' 'Broadcast engineer'

'Chartered management accountant' 'Theatre manager' 'Animal technologist'
 'Animator' 'Producer, radio' 'Chiropodist' 'Exhibition designer'
 'Occupational therapist' 'Database administrator'
 'Arts development officer' 'Health and safety inspector'
 'Press photographer' 'Recruitment consultant'
 'Dance movement psychotherapist' 'Audiological scientist'
 'Soil scientist' 'Equities trader' 'Orthoptist' 'Engineer, materials'
 'Regulatory affairs officer' 'Trade union research officer'
 'Research scientist (maths)' 'Television production assistant'
 'Chief of Staff' 'Advertising copywriter'
 'Programme researcher, broadcasting/film/video'
 'Technical sales engineer' 'Music therapist' 'Electronics engineer'
 'Waste management officer' 'Plant breeder/geneticist'
 'Operational researcher' 'Further education lecturer'
 'Electrical engineer' 'Television camera operator'
 'Runner, broadcasting/film/video' 'Pharmacist, community'
 'Ophthalmologist' 'Wellsite geologist' 'Psychologist, educational'
 'Advertising account planner' 'Sports therapist'
 'Surveyor, building control' 'Engineer, land' 'Clinical embryologist'
 'Marine scientist' 'Teacher, secondary school' 'Chief Financial Officer'
 'Landscape architect' 'Community pharmacist' 'Product manager'
 'Financial risk analyst' 'Administrator' 'Civil engineer, contracting'
 'Engineer, maintenance (IT)' 'Scientist, audiological'
 'Management consultant' 'Dentist' 'Barrister' 'Surveyor, insurance'
 'Customer service manager' 'Clinical cytogeneticist'
 'Forest/woodland manager' 'Insurance underwriter'
 'Speech and language therapist' 'Trading standards officer'
 'Surveyor, building' 'Engineering geologist' 'Investment analyst'
 'Research scientist (life sciences)' 'Firefighter'
 'Higher education careers adviser' 'Theatre director'
 'Passenger transport manager' 'English as a foreign language teacher'
 'Research officer, trade union'
 'Conservation officer, historic buildings'
 'Scientist, product/process development' 'Air cabin crew'
 'Colour technologist' 'Research officer, political party'
 'Chemist, analytical' 'Hydrogeologist' 'Music tutor' 'Therapist, drama'
 'Health physicist' 'Lecturer, higher education' 'Records manager'
 'Scientist, research (medical)' 'Field trials officer'
 'Adult guidance worker' 'Fine artist'
 'Social research officer, government' 'Interior and spatial designer'
 'Freight forwarder' 'Production engineer' 'Accommodation manager'
 'Retail banker' 'Research scientist (medical)' 'Occupational hygienist'
 'Diplomatic Services operational officer' 'Barrister's clerk'
 'Call centre manager' 'Tourism officer' 'Agricultural consultant'
 'Armed forces technical officer' 'Politician's assistant'
 'Geographical information systems officer' 'Chief Operating Officer'
 'Higher education lecturer' 'Therapist, occupational' 'Land'
 'Print production planner' 'Tree surgeon' 'Physiological scientist'

'Producer, television/film/video' 'Facilities manager'
 'Designer, blown glass/stained glass' 'Location manager'
 'Maintenance engineer' 'Meteorologist' 'Local government officer'
 'Energy manager' 'Estate agent' 'Counsellor' 'Dispensing optician'
 'Geophysical data processor' 'Adult nurse' 'Educational psychologist'
 'Mental health nurse' 'IT sales professional' 'Water quality scientist'
 'Advice worker' 'Intelligence analyst' 'Community arts worker'
 'Optometrist' 'Patent examiner' 'Psychotherapist, dance movement'
 'Gaffer' 'Risk analyst' 'Financial trader'
 'Sales promotion account executive' 'Equality and diversity officer'
 'Administrator, education' 'Medical secretary'
 'Claims inspector/assessor' 'Child psychotherapist' 'Immigration officer'
 'Metallurgist' 'Education administrator' 'Fitness centre manager'
 'Chief Strategy Officer' 'Public librarian'
 'Furniture conservator/restorer' 'Photographer' 'Production manager'
 'Nature conservation officer' 'Phytotherapist' 'Therapist, horticultural'
 'Aeronautical engineer' 'Engineer, civil (consulting)'
 'Television/film/video producer' 'Solicitor, Scotland'
 'Psychologist, forensic' 'Development worker, community'
 'Engineer, manufacturing' 'Garment/textile technologist'
 'Charity officer' 'Insurance risk surveyor' 'Broadcast presenter'
 'Secretary/administrator' 'Civil Service administrator'
 'Surveyor, hydrographic' 'Loss adjuster, chartered'
 'Secondary school teacher' 'Teacher, special educational needs'
 'Engineer, petroleum' 'Surveyor, rural practice'
 'Information systems manager' 'Designer, furniture' 'Engineer, energy'
 'Conservator, museum/gallery' 'Environmental consultant'
 'Doctor, general practice' 'Nurse, mental health' 'Graphic designer'
 'Investment banker, corporate' 'Astronomer' 'Data processing manager'
 'Stage manager' 'Textile designer' 'Drilling engineer'
 'Scientist, research (life sciences)' 'Furniture designer'
 'Ambulance person' 'Buyer, industrial' 'Copywriter, advertising'
 'Academic librarian' 'Scientist, research (maths)'
 'International aid/development worker' 'Engineer, structural'
 'Lecturer, further education' 'Interpreter' 'Chief Marketing Officer'
 'Transport planner' 'Pharmacist, hospital' 'Toxicologist' 'Proofreader'
 'Contracting civil engineer' 'Psychologist, clinical' 'Retail manager'
 'Manufacturing systems engineer' 'Art therapist'
 'Chartered certified accountant' 'Sales professional, IT'
 'Dramatherapist' 'Designer, interior/spatial'
 'Administrator, Civil Service' 'Printmaker' 'Engineer, electrical'
 'Planning and development surveyor' 'Paediatric nurse'
 'Designer, multimedia' 'Herpetologist' 'Mudlogger' 'Engineer, water'
 'Arboriculturist' 'Sub' 'Sports administrator' 'Mechanical engineer'
 'Physicist, medical' 'Armed forces training and education officer'
 'Marketing executive' 'Magazine features editor' 'Ergonomist'
 'Mining engineer' 'Dancer' 'Optician, dispensing' 'Designer, textile'
 'Ranger/warden' 'Psychiatrist' 'Bonds trader' 'Technical brewer'

'Engineer, building services' 'Field seismologist'
 'Engineer, electronics' 'Medical illustrator' 'Architect'
 'Engineer, production' 'Licensed conveyancer' 'Surveyor, mining'
 'Applications developer' 'Museum/gallery curator' 'Market researcher'
 'Radiation protection practitioner'
 'Control and instrumentation engineer' 'Programmer, applications'
 'Advertising art director'
 'Clinical scientist, histocompatibility and immunogenetics'
 'Professor Emeritus' 'Horticulturist, amenity' 'Physiotherapist'
 'Race relations officer' 'Surveyor, land/geomatics' 'Youth worker'
 'Horticultural therapist' 'IT consultant' 'Make'
 'Public relations account executive' 'Private music teacher'
 'Fashion designer' 'Hospital pharmacist' 'Tax adviser'
 'Engineer, broadcasting (operations)' 'Commercial art gallery manager'
 'Legal executive' 'Visual merchandiser' 'Commercial/residential surveyor'
 'Personal assistant' 'Insurance claims handler' 'Financial manager'
 'Tourist information centre manager' 'Scientist, physiological'
 'Designer, ceramics/pottery' 'Accountant, chartered management'
 'Psychotherapist' 'Health visitor' 'Pharmacologist'
 'Special educational needs teacher' 'Public relations officer'
 'Town planner' 'Animal nutritionist' 'Building control surveyor'
 'Engineer, automotive' 'Information officer'
 'Senior tax professional/tax inspector' 'Film/video editor' 'Cabin crew'
 'Radiographer, diagnostic' 'Warden/ranger' 'Video editor' 'Airline pilot'
 'Newspaper journalist' 'Education officer, community'
 'Geologist, engineering' 'Librarian, academic' 'Paramedic'
 'Recycling officer' 'Merchandiser, retail' 'Retail merchandiser'
 'Administrator, local government' 'Counselling psychologist'
 'Estate manager/land agent' 'Oceanographer' 'Haematologist'
 'Scientist, research (physical sciences)' 'Medical physicist'
 'Communications engineer' 'Surgeon' 'Homeopath' 'Charity fundraiser'
 'Theme park manager' 'Barista' 'Chartered public finance accountant'
 'Teaching laboratory technician' 'Microbiologist'
 'Programmer, multimedia' 'Automotive engineer' 'Holiday representative'
 'Systems analyst' 'Product designer' 'Forensic scientist'
 'Museum/gallery conservator' 'Patent attorney' 'Ship broker'
 'Technical author' 'Pension scheme manager' 'Ceramics designer'
 'Careers adviser' 'Building surveyor' 'Public house manager'
 'Environmental education officer' 'Journalist, magazine'
 'Magazine journalist' 'Analytical chemist'
 'Teacher, English as a foreign language'
 'Lighting technician, broadcasting/film/video' 'Teacher, early years/pre'
 'Commercial horticulturist' 'Publishing copy' 'Clinical biochemist'
 'IT trainer' 'Programmer, systems' 'Logistics and distribution manager'
 'Horticultural consultant' 'Hotel manager' 'Associate Professor'
 'Nurse, learning disability' 'Hydrographic surveyor' 'Nurse, adult'
 'Fisheries officer' 'Administrator, sports' 'Insurance broker'
 'Veterinary surgeon' 'Designer, jewellery' 'Lobbyist' 'Chemical engineer'

```

'Chartered loss adjuster' 'Social researcher' 'Petroleum engineer'
'Social worker' 'Education officer, environmental' 'Futures trader'
'Fish farm manager' 'Lawyer' 'Seismic interpreter' 'TEFL teacher'
'Immunologist' 'Engineer, drilling'
'Emergency planning/management officer' 'Pathologist'
'Broadcast journalist' 'Geologist, wellsite'
'Investment banker, operational' 'Biomedical scientist' 'Bookseller'
'Copy' 'Midwife' 'Media buyer' 'Geneticist, molecular'
'Housing manager/officer' 'Geophysicist/field seismologist'
'Art gallery manager' 'Food technologist' 'Land/geomatics surveyor'
'Radio broadcast assistant' 'Psychologist, prison and probation services'
'Dietitian' 'Civil engineer, consulting' 'Sales executive'
'Leisure centre manager' 'Scientist, biomedical'
'Exhibitions officer, museum/gallery' 'Engineer, communications'
'Catering manager' 'Administrator, arts' 'Software engineer'
'Medical laboratory scientific officer' 'Commissioning editor'
'Geoscientist' 'Materials engineer' 'Financial planner'
'Brewing technologist' 'Minerals surveyor' 'Editor, magazine features'
'General practice doctor' 'Health and safety adviser' 'Doctor, hospital'
'Environmental manager' 'Clinical research associate'
'Sound technician, broadcasting/film/video' 'Press sub' 'Retail buyer'
'Comptroller' 'Government social research officer'
'Rural practice surveyor' 'Accountant, chartered']
Children, float64 :
[ 1.  3.  0.  7.  2.  4. 10.  5.  6.  9.  8.]
Age, int32 :
[53 51 78 22 76 50 40 48 55 64 41 45 85 44 54 72 84 68 52 31 60 75 70 63
 56 32 86 65 66 67 79 25 58 59 33 83 73 43 57 36 49 39 20 69 26 47 18 38
 82 34 74 37 77 27 89 30 87 23 29 80 19 24 88 62 46 71 21 61 81 42 35 28]
Education, object :
['Some College, Less than 1 Year'
'Some College, 1 or More Years, No Degree'
'GED or Alternative Credential' 'Regular High School Diploma'
"Bachelor's Degree" "Master's Degree" 'Nursery School to 8th Grade'
'9th Grade to 12th Grade, No Diploma' 'Doctorate Degree'
"Associate's Degree" 'Professional School Degree'
'No Schooling Completed']
Employment, object :
['Full Time' 'Retired' 'Unemployed' 'Student' 'Part Time']
Income, float64 :
[86575.93 46805.99 14370.14 ... 65917.81 29702.32 62682.63]
Marriage_status, object :
['Divorced' 'Married' 'Widowed' 'Never Married' 'Separated']
Gender, object :
['Male' 'Female' 'Prefer not to answer']
Readmitted, object :
['No' 'Yes']
VitD_levels, float64 :

```

```

[17.80233049 18.99463952 17.4158887 ... 15.75275136 21.95630508
 20.42188348]
Doc_visits, int64 :
[6 4 5 7 3 2 8 9 1]
Full_meals_eaten, int64 :
[0 2 1 3 4 5 7 6]
VitD_supplements, int64 :
[0 1 2 3 4 5]
Habitual_soft_drink_use, object :
['No' 'Yes']
Initial_admin, object :
['Emergency Admission' 'Elective Admission' 'Observation Admission']
High_blood_pressure, object :
['Yes' 'No']
Stroke, object :
['No' 'Yes']
Complication_risk, object :
['Medium' 'High' 'Low']
Overweight, float64 :
[0. 1.]
Arthritis, object :
['Yes' 'No']
Diabetes, object :
['Yes' 'No']
Hyperlipidemia, object :
['No' 'Yes']
Back_pain, object :
['Yes' 'No']
Anxiety, float64 :
[1. 0.]
Allergic_rhinitis, object :
['Yes' 'No']
Reflux_esophagitis, object :
['No' 'Yes']
Asthma, object :
['Yes' 'No']
Primary_service_recived, object :
['Blood Work' 'Intravenous' 'CT Scan' 'MRI']
Initial_days, float64 :
[10.58576971 15.12956221 4.77217721 ... 66.01257016 63.35690285
 70.85059182]
Total_charge, float64 :
[3191.048774 4214.905346 2177.586768 ... 7725.953391 8462.831883
 8700.856021]
Additional_charges, float64 :
[17939.40342 17612.99812 17505.19246 ... 15281.21466 7781.678412
 11643.18993 ]
Survey_timely_addmission, int64 :

```

```

[3 2 4 1 5 7 6 8]
Survey_timely_treatment, int64 :
[3 4 5 1 2 6 7]
Survey_timely_visits, int64 :
[2 3 4 5 1 6 7 8]
Survey_reliability, int64 :
[2 4 3 5 6 1 7]
Survey_options, int64 :
[4 3 5 2 6 1 7]
Survey_hours, int64 :
[3 4 5 2 6 1 7]
Survey_courtesy, int64 :
[3 5 4 2 6 1 7]
Survey_active_listening, int64 :
[4 3 5 6 2 1 7]

```

```

[39]: #standardize time zones to utc, if a time zone does not observe daylight
      ↪ savings time it is appended with (ND)
timezone_dict = {'America/Chicago': 'UTC-6:00', 'America/New_York': 'UTC-5:00',
      ↪ 'America/Los_Angeles': 'UTC-8:00',
      'America/Indiana/Indianapolis': 'UTC-5:00', 'America/Detroit': 'UTC-5:00',
      ↪ 'America/Denver': 'UTC-7:00',
      'America/Nome': 'UTC-9:00', 'America/Anchorage': 'UTC-9:00', 'America/
      ↪ Phoenix': 'UTC-8:00(ND)',
      'America/Boise': 'UTC-8:00', 'America/Puerto_Rico': 'UTC-4:00(ND)', 'America/
      ↪ Yakutat': 'UTC-9:00',
      'Pacific/Honolulu': 'UTC-10:00(ND)', 'America/Menominee': 'UTC-6:00', 'America/
      ↪ Kentucky/Louisville': 'UTC-5:00',
      'America/Indiana/Vincennes': 'UTC-5:00', 'America/Toronto': 'UTC-5:00',
      ↪ 'America/Indiana/Marengo': 'UTC-5:00',
      'America/Indiana/Winamac': 'UTC-5:00', 'America/Indiana/Tell_City': 'UTC-6:
      ↪ 00', 'America/Sitka': 'UTC-9:00',
      'America/Indiana/Knox': 'UTC-6:00', 'America/North_Dakota/New_Salem': 'UTC-6:
      ↪ 00', 'America/Indiana/Vevay': 'UTC-5:00',
      'America/Adak': 'UTC-10:00', 'America/North_Dakota/Beulah': 'UTC-6:00'}
df['Timezone'].replace(timezone_dict, inplace = True)
df['Timezone']

```

```

[39]: Case_order
1      UTC-6:00
2      UTC-6:00
3      UTC-6:00
4      UTC-6:00
5      UTC-5:00
...
9996   UTC-5:00
9997   UTC-5:00

```

```

9998      UTC-6:00
9999      UTC-7:00
10000     UTC-5:00
Name: Timezone, Length: 10000, dtype: object

```

```

[40]: #Convert columns that that express whole numer values that are currently float
      ↪ to int
df.loc[:,['Children', 'Overweight', 'Anxiety']] = df[['Children', 'Overweight',
      ↪ 'Anxiety']].astype(int)

```

```

[41]: #convert zip colum to string type, identify records with invalid zip codes
df['Zip'] = df['Zip'].astype(str)

```

```

[42]: invalid_zips = df['Zip'].apply(len) != 5
      invalid_list = df.loc[invalid_zips, ['Zip', 'City', 'State']]
      invalid_list

```

```

[42]:
      Zip      City State
Case_order
32      2584      Nantucket  MA
36      5043      East Thetford  VT
37      2468      Waban  MA
38      2138      Cambridge  MA
68      3464      Stoddard  NH
...
9976     4415  Brownville Junction  ME
9977     6084      Tolland  CT
9983     8401      Atlantic City  NJ
9994     7647      Northvale  NJ
9997     8340      Milmay  NJ

```

[723 rows x 3 columns]

```

[43]: #list all invalid zip codes, cities, and states in list
      for i in range(0, invalid_list.shape[0]):
          print(invalid_list.iloc[i])
      #while manually cross referencing this data against a
      #United states zip code database(https://www.zipdatamaps.com/index.php),
      #it became apparent that the invalid zip codes where caused by leading 0's
      ↪ being ommited

```

```

Zip      2584
City      Nantucket
State      MA
Name: 32, dtype: object
Zip      5043
City      East Thetford
State      VT

```

Name: 36, dtype: object
 Zip 2468
 City Waban
 State MA
 Name: 37, dtype: object
 Zip 2138
 City Cambridge
 State MA
 Name: 38, dtype: object
 Zip 3464
 City Stoddard
 State NH
 Name: 68, dtype: object
 Zip 8332
 City Millville
 State NJ
 Name: 109, dtype: object
 Zip 7935
 City Green Village
 State NJ
 Name: 114, dtype: object
 Zip 7882
 City Washington
 State NJ
 Name: 120, dtype: object
 Zip 3462
 City Spofford
 State NH
 Name: 145, dtype: object
 Zip 4408
 City Aurora
 State ME
 Name: 149, dtype: object
 Zip 4940
 City Farmington Falls
 State ME
 Name: 172, dtype: object
 Zip 2889
 City Warwick
 State RI
 Name: 174, dtype: object
 Zip 3885
 City Stratham
 State NH
 Name: 190, dtype: object
 Zip 2835
 City Jamestown
 State RI

Name: 195, dtype: object
 Zip 3220
 City Belmont
 State NH
 Name: 198, dtype: object
 Zip 4344
 City Farmingdale
 State ME
 Name: 203, dtype: object
 Zip 669
 City Lares
 State PR
 Name: 226, dtype: object
 Zip 7030
 City Hoboken
 State NJ
 Name: 248, dtype: object
 Zip 4449
 City Hudson
 State ME
 Name: 309, dtype: object
 Zip 7630
 City Emerson
 State NJ
 Name: 310, dtype: object
 Zip 6119
 City West Hartford
 State CT
 Name: 311, dtype: object
 Zip 3446
 City Swanzey
 State NH
 Name: 313, dtype: object
 Zip 4926
 City China Village
 State ME
 Name: 338, dtype: object
 Zip 4344
 City Farmingdale
 State ME
 Name: 341, dtype: object
 Zip 2364
 City Kingston
 State MA
 Name: 345, dtype: object
 Zip 6401
 City Ansonia
 State CT

Name: 368, dtype: object
 Zip 8876
 City Somerville
 State NJ
 Name: 388, dtype: object
 Zip 6263
 City Rogers
 State CT
 Name: 393, dtype: object
 Zip 4626
 City Cutler
 State ME
 Name: 395, dtype: object
 Zip 7028
 City Glen Ridge
 State NJ
 Name: 416, dtype: object
 Zip 6498
 City Westbrook
 State CT
 Name: 417, dtype: object
 Zip 6264
 City Scotland
 State CT
 Name: 457, dtype: object
 Zip 8004
 City Atco
 State NJ
 Name: 474, dtype: object
 Zip 2838
 City Manville
 State RI
 Name: 486, dtype: object
 Zip 4530
 City Bath
 State ME
 Name: 512, dtype: object
 Zip 1832
 City Haverhill
 State MA
 Name: 513, dtype: object
 Zip 7716
 City Atlantic Highlands
 State NJ
 Name: 522, dtype: object
 Zip 7460
 City Stockholm
 State NJ

Name: 524, dtype: object
 Zip 1940
 City Lynnfield
 State MA
 Name: 530, dtype: object
 Zip 7055
 City Passaic
 State NJ
 Name: 532, dtype: object
 Zip 4988
 City Unity
 State ME
 Name: 533, dtype: object
 Zip 1562
 City Spencer
 State MA
 Name: 534, dtype: object
 Zip 6375
 City Quaker Hill
 State CT
 Name: 542, dtype: object
 Zip 3227
 City Center Sandwich
 State NH
 Name: 546, dtype: object
 Zip 7311
 City Jersey City
 State NJ
 Name: 552, dtype: object
 Zip 2745
 City New Bedford
 State MA
 Name: 561, dtype: object
 Zip 7857
 City Netcong
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 Name: 9983, dtype: object
 Zip 7647
 City Northvale
 State NJ

```
Name: 9994, dtype: object
Zip      8340
City     Milmay
State    NJ
Name: 9997, dtype: object
```

```
[44]: #correct invalid zipcodes
invalid_zip_indexes = invalid_list.index.values
for x in invalid_zip_indexes:
    df.loc[x, 'Zip'] = df.loc[x, 'Zip'].zfill(5)
```

```
[45]: df['Zip'][df['Zip'].apply(len) != 5]
```

```
[45]: Series([], Name: Zip, dtype: object)
```

```
[46]: df.loc[invalid_zip_indexes, 'Zip']
```

```
[46]: Case_order
32      02584
36      05043
37      02468
38      02138
68      03464
...
9976    04415
9977    06084
9983    08401
9994    07647
9997    08340
Name: Zip, Length: 723, dtype: object
```

```
[47]: #Round total and additional charges to 2 decimal places. These values were
↳ generated based on averages
#and were not standardized for typical use of monetary values
df[['Total_charge', 'Additional_charges']] = np.around(df[['Total_charge',
↳ 'Additional_charges']], 2)
df[['Total_charge', 'Additional_charges']]
```

```
[47]:
```

	Total_charge	Additional_charges
Case_order		
1	3191.05	17939.40
2	4214.91	17613.00
3	2177.59	17505.19
4	2465.12	12993.44
5	1885.66	3716.53
...
9996	6651.24	8927.64
9997	7851.52	28507.15

9998	7725.95	15281.21
9999	8462.83	7781.68
10000	8700.86	11643.19

[10000 rows x 2 columns]

```
[48]: #reduce precision of initial days variable to allow for more meaningful data
      ↪analysis
df[['Initial_days']] = np.around(df[['Initial_days']], 1)
df['Initial_days'].value_counts()
```

```
[48]: 3.3      56
      1.3      48
      7.8      47
      2.8      44
      1.6      43
      ..
      22.7     1
      32.1     1
      28.9     1
      25.0     1
      31.8     1
Name: Initial_days, Length: 646, dtype: int64
```

```
[49]: #isolate numeric values for outlier detection
numeric_data = df[['Population', 'Children', 'Age', 'Income', 'VitD_levels',
      ↪'Doc_visits', 'Full_meals_eaten',
      'VitD_supplements', 'Initial_days', 'Total_charge',
      ↪'Additional_charges']].copy()
```

Outliers are identified and isolated using a combination of box plots and z scores. Where needed histograms are used for further analysis. In cases where z scores were not suitable for outlier isolation iqr was used instead. outliers are stored in a separate variable named `__outliers`, but not removed from the original dataset. This is done so that analysis can be performed on dataset both including and excluding outliers, because while outliers are present they are not abnormal values for the data type.

```
[50]: #helper function to add boolean outlier column to main dataframe for a specific
      ↪column. this can be used during later
#data analysis to easily include or exclude outliers from analysis
def Add_outlier_column(data_frame, outliers, column):
    data_frame[column + '_outliers'] = False
    for x in outliers.index:
        data_frame.at[x-1, column + '_outliers'] = True
```

```
[51]: for x in numeric_data:
        numeric_data[x + '_z'] = stats.zscore(numeric_data[x])
numeric_data
```

```
[51]:
```

	Population	Children	Age	Income	VitD_levels	Doc_visits	\
Case_order							
1	2951	1	53	86575.93	17.802330	6	
2	11303	3	51	46805.99	18.994640	4	
3	17125	3	53	14370.14	17.415889	4	
4	2162	0	78	39741.49	17.420079	4	
5	5287	0	22	1209.56	16.870524	5	
...	
9996	4762	6	25	45967.61	16.481612	4	
9997	1251	4	87	14983.02	18.451601	5	
9998	532	3	65	65917.81	15.752751	4	
9999	271	3	43	29702.32	21.956305	5	
10000	41524	8	43	62682.63	20.421883	5	

	Full_meals_eaten	VitD_supplements	Initial_days	Total_charge	\
Case_order					
1	0	0	10.6	3191.05	
2	2	1	15.1	4214.91	
3	1	0	4.8	2177.59	
4	1	0	1.7	2465.12	
5	0	2	1.3	1885.66	
...	
9996	2	1	51.6	6651.24	
9997	0	0	68.7	7851.52	
9998	2	0	66.0	7725.95	
9999	2	1	63.4	8462.83	
10000	0	1	70.9	8700.86	

	...	Children_z	Age_z	Income_z	VitD_levels_z	Doc_visits_z	\
Case_order	...						
1	...	-0.510287	-0.014419	1.709132	-0.239530	0.944647	
2	...	0.412224	-0.117337	0.233169	-0.062181	-0.967981	
3	...	0.412224	-0.014419	-0.970607	-0.297011	-0.967981	
4	...	-0.971543	1.272056	-0.029012	-0.296388	-0.967981	
5	...	-0.971543	-1.609647	-1.459030	-0.378131	-0.011667	
...	
9996	...	1.795992	-1.455270	0.202055	-0.435979	-0.967981	
9997	...	0.873480	1.735187	-0.947862	-0.142954	-0.011667	
9998	...	0.412224	0.603089	0.942457	-0.544393	-0.967981	
9999	...	0.412224	-0.529009	-0.401591	0.378351	-0.011667	
10000	...	2.718503	-0.529009	0.822391	0.150114	-0.011667	

	Full_meals_eaten_z	VitD_supplements_z	Initial_days_z	\
Case_order				
1	-0.993387	-0.634713	-0.908650	
2	0.990609	0.956445	-0.737310	
3	-0.001389	-0.634713	-1.129488	

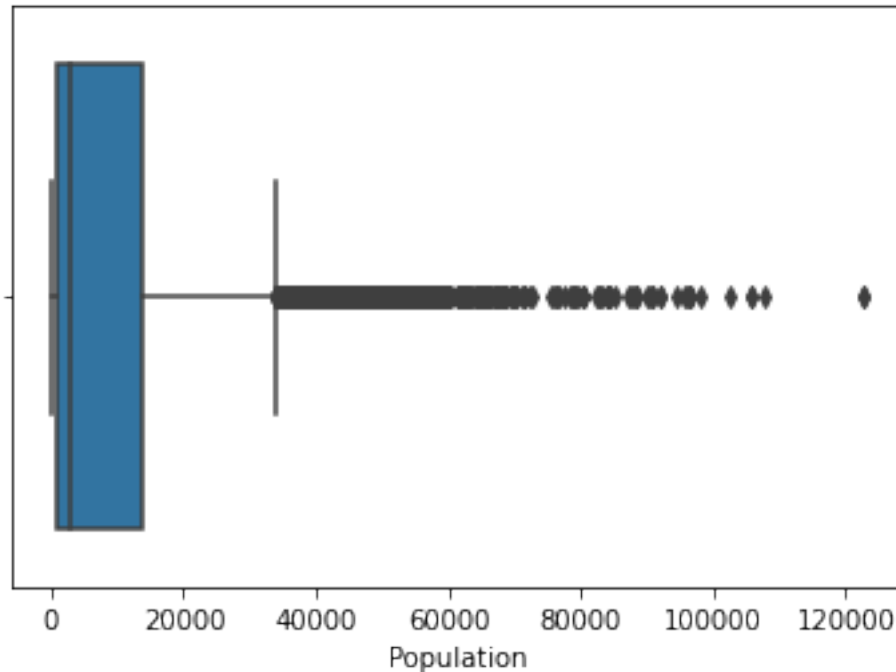
4	-0.001389	-0.634713	-1.247522
5	-0.993387	2.547602	-1.262752
...
9996	0.990609	0.956445	0.652447
9997	-0.993387	-0.634713	1.303539
9998	0.990609	-0.634713	1.200735
9999	0.990609	0.956445	1.101738
10000	-0.993387	0.956445	1.387305

Case_order	Total_charge_z	Additional_charges_z
1	-0.799579	0.765005
2	-0.496427	0.715114
3	-1.099651	0.698635
4	-1.014517	0.009005
5	-1.186087	-1.408990
...
9996	0.224938	-0.612461
9997	0.580324	2.380307
9998	0.543145	0.358695
9999	0.761325	-0.787623
10000	0.831803	-0.197384

[10000 rows x 22 columns]

```
[52]: sns.boxplot(x=numeric_data['Population'])
```

```
[52]: <AxesSubplot:xlabel='Population'>
```



```
[53]: population_outliers = numeric_data.loc[(numeric_data['Population_z'] > 3) |
      ↪(numeric_data['Population_z'] < -3),
      ['Population', 'Population_z']]
Add_outlier_column(df, population_outliers, 'Population')
population_outliers.sort_values('Population')
```

```
[53]:
```

	Population	Population_z
Case_order		
289	54453	3.001059
965	54460	3.001531
6797	54507	3.004701
3820	54647	3.014146
3186	54647	3.014146
...
768	105799	6.464762
7687	105799	6.464762
5966	107700	6.593000
9663	122814	7.612562
3025	122814	7.612562

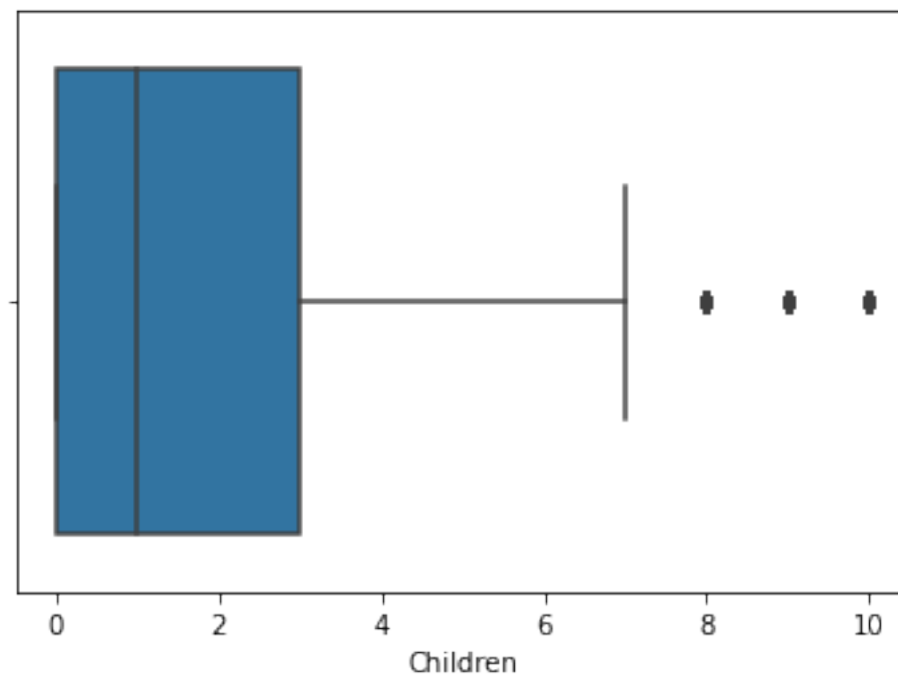
[218 rows x 2 columns]

```
[54]: population_outliers.sort_values('Population').value_counts()
```

```
[54]: Population  Population_z
      57775      3.225154      3
      83960      4.991545      3
      67597      3.887728      3
      59129      3.316493      2
      84418      5.022441      2
      ..
      59699      3.354944      1
      60033      3.377475      1
      60081      3.380713      1
      60107      3.382467      1
      63425      3.606293      1
      Length: 186, dtype: int64
```

```
[55]: sns.boxplot(x=numeric_data['Children'])
```

```
[55]: <AxesSubplot:xlabel='Children'>
```



```
[56]: children_outliers = numeric_data.loc[(numeric_data['Children_z'] > 3) |
      ↪(numeric_data['Children_z'] < -3),
      ['Children', 'Children_z']]
Add_outlier_column(df, children_outliers, 'Children')
children_outliers.sort_values('Children')
```

```
[56]:
```

	Children	Children_z
Case_order		
4459	9	3.179759
4134	9	3.179759
4110	9	3.179759
4049	9	3.179759
4048	9	3.179759
...
2282	10	3.641015
2196	10	3.641015
2125	10	3.641015
6831	10	3.641015
9846	10	3.641015

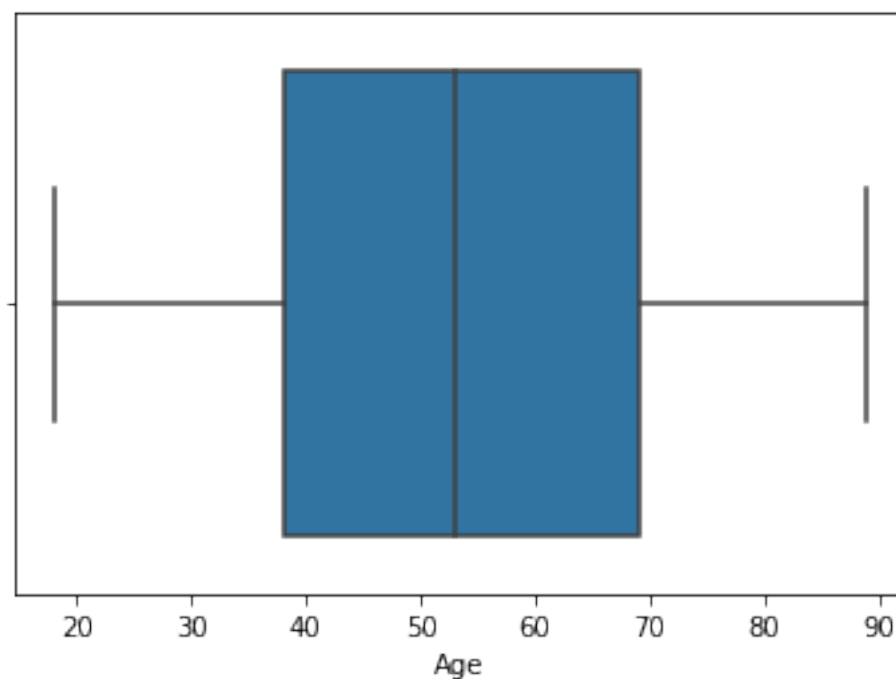
[210 rows x 2 columns]

```
[57]: children_outliers.sort_values('Children').value_counts()
```

```
[57]: Children  Children_z
9          3.179759      125
10         3.641015       85
dtype: int64
```

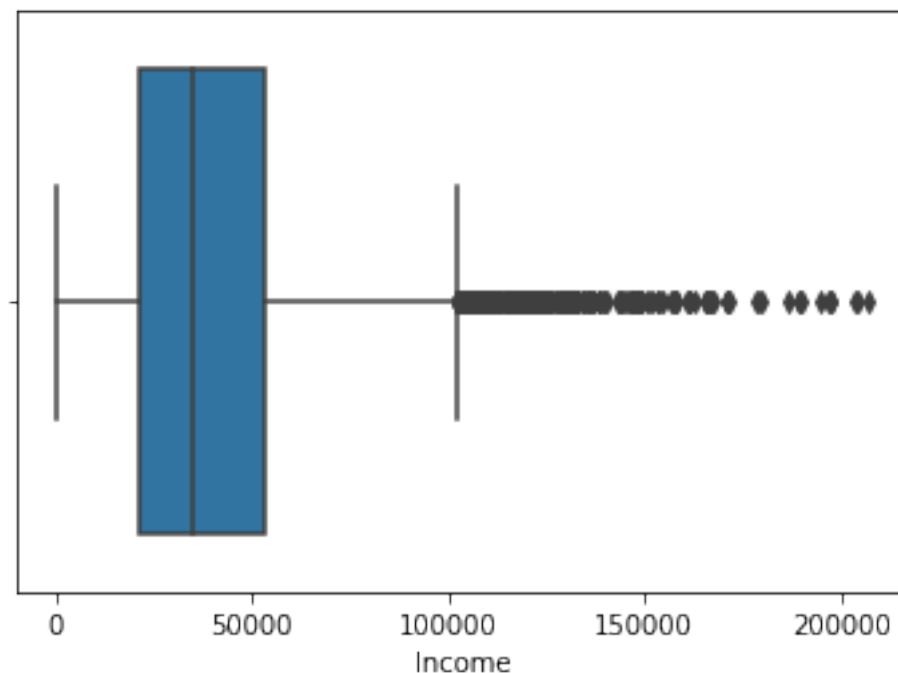
```
[58]: sns.boxplot(x=numeric_data['Age'])
```

```
[58]: <AxesSubplot:xlabel='Age'>
```




```
[59]: sns.boxplot(x=numeric_data['Income'])
```

```
[59]: <AxesSubplot:xlabel='Income'>
```



```
[60]: income_outliers = numeric_data.loc[(numeric_data['Income_z'] > 3) |
↳(numeric_data['Income_z'] < -3),
      ['Income', 'Income_z']]
Add_outlier_column(df, income_outliers, 'Income')
income_outliers.sort_values('Income')
```

```
[60]:
```

	Income	Income_z
Case_order		
1515	121766.35	3.015137
9345	121931.19	3.021255
9956	122291.51	3.034627
9141	122361.47	3.037224
37	122615.82	3.046663
...
1779	197576.18	5.828632
6407	197675.05	5.832301
8599	203774.65	6.058672
842	204542.41	6.087166
8387	207249.13	6.187619

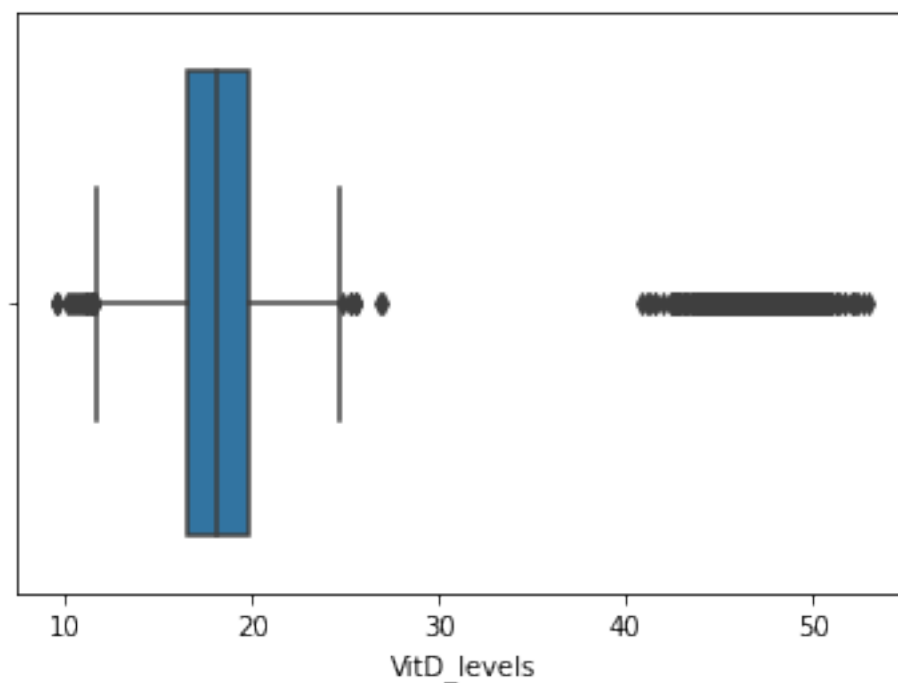
[140 rows x 2 columns]

```
[61]: income_outliers.sort_values('Income').value_counts()
```

```
[61]: Income      Income_z
121766.35  3.015137      1
148944.14  4.023774      1
147303.68  3.962892      1
147570.86  3.972808      1
148141.83  3.993998      1
..
129987.32  3.320238      1
129945.51  3.318687      1
129586.68  3.305370      1
129349.07  3.296551      1
207249.13  6.187619      1
Length: 140, dtype: int64
```

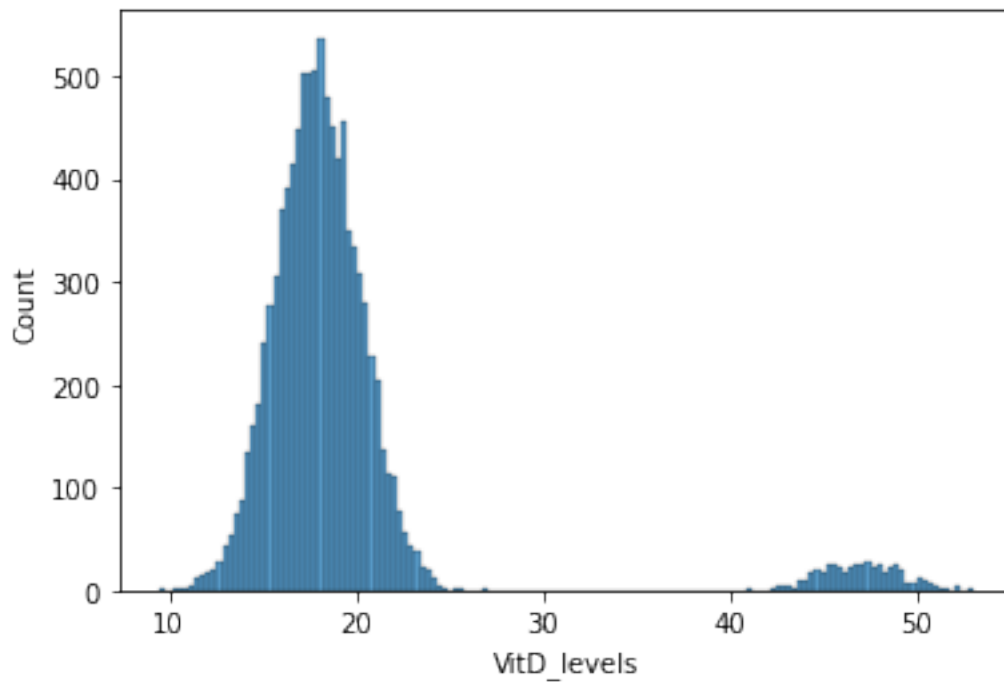
```
[62]: sns.boxplot(x=numeric_data['VitD_levels'])
```

```
[62]: <AxesSubplot:xlabel='VitD_levels'>
```



```
[63]: sns.histplot(numeric_data['VitD_levels'])
```

```
[63]: <AxesSubplot:xlabel='VitD_levels', ylabel='Count'>
```

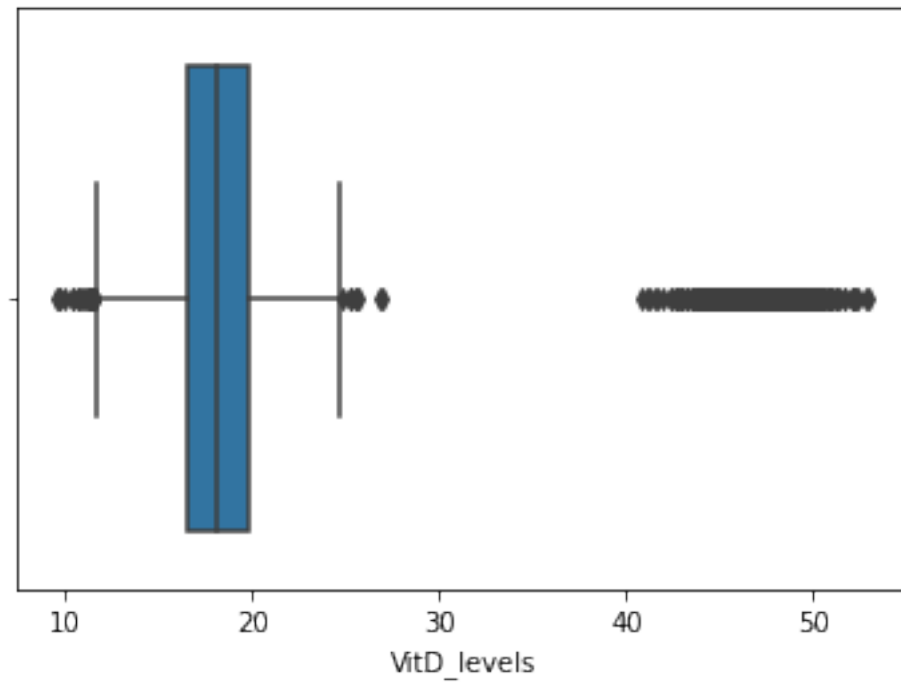


VitD column in both numeric data and full data frame precision reduced to tenths decimal place to conform to typical measurement of data of this type as shown in the following examples: <https://www.questdiagnostics.com/home/physicians/testing-services/condition/endocrinology/what-low-vitamin-d-numbers-mean/>
<https://ods.od.nih.gov/factsheets/Vitamin%20D-HealthProfessional/>
<https://www.health.harvard.edu/blog/vitamin-d-whats-right-level-2016121910893>

```
[64]: numeric_data[['VitD_levels']] = df[['VitD_levels']] = np.  
      ↪around(numeric_data['VitD_levels'], 1)  
      numeric_data['VitD_levels_z'] = stats.zscore(numeric_data['VitD_levels'])
```

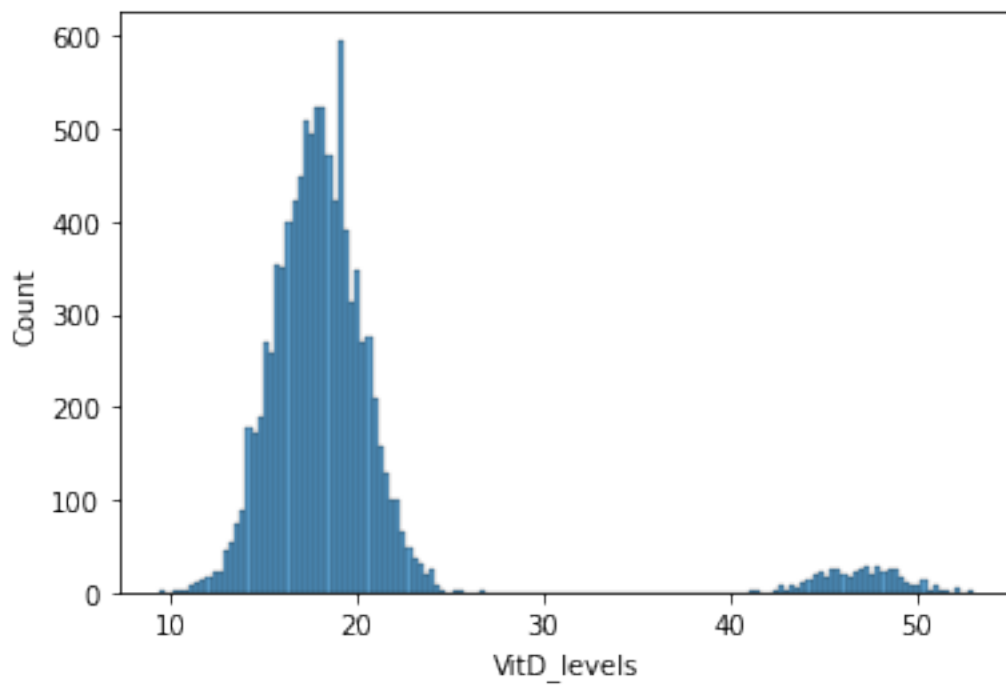
```
[65]: sns.boxplot(x=numeric_data['VitD_levels'])
```

```
[65]: <AxesSubplot:xlabel='VitD_levels'>
```



```
[66]: sns.histplot(numeric_data['VitD_levels'])
```

```
[66]: <AxesSubplot:xlabel='VitD_levels', ylabel='Count'>
```



```
[67]: vitD_levels_outliers = numeric_data.loc[(numeric_data['VitD_levels_z'] > 3) |
↳(numeric_data['VitD_levels_z'] < -3),
      ['VitD_levels', 'VitD_levels_z']]
Add_outlier_column(df, vitD_levels_outliers, 'VitD_levels')
vitD_levels_outliers.sort_values('VitD_levels')
```

```
[67]:      VitD_levels  VitD_levels_z
Case_order
8198           40.8         3.181200
787            41.1         3.225822
7271           41.2         3.240697
2947           41.5         3.285319
5689           41.6         3.300193
...
2616           52.2         4.876861
7231           52.3         4.891736
7158           52.4         4.906610
1307           52.8         4.966107
1964           53.0         4.995855
```

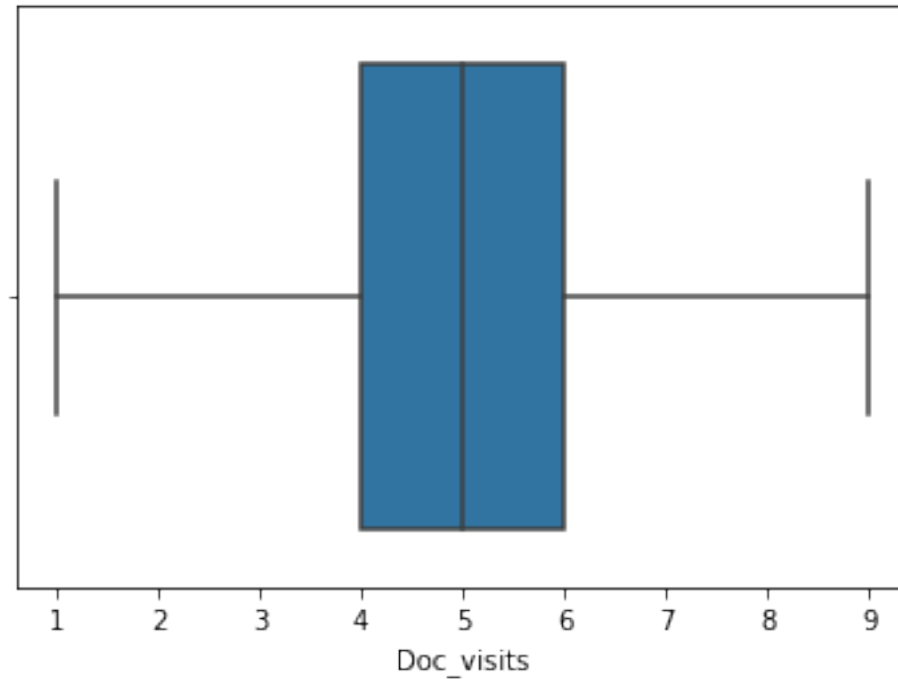
[500 rows x 2 columns]

```
[68]: vitD_levels_outliers.sort_values('VitD_levels').value_counts()
```

```
[68]: VitD_levels  VitD_levels_z
47.8          4.222395         12
46.6          4.043905         11
48.5          4.326515         11
45.6          3.895162         11
45.9          3.939785         11
..
42.0          3.359690          1
41.6          3.300193          1
41.5          3.285319          1
41.2          3.240697          1
53.0          4.995855          1
Length: 99, dtype: int64
```

```
[69]: sns.boxplot(x=numeric_data['Doc_visits'])
```

```
[69]: <AxesSubplot:xlabel='Doc_visits'>
```



```
[70]: doc_visits_outliers = numeric_data.loc[(numeric_data['Doc_visits_z'] > 3) |
      ↪(numeric_data['Doc_visits_z'] < -3),
      ['Doc_visits', 'Doc_visits_z']]
Add_outlier_column(df, doc_visits_outliers, 'Doc_visits')
doc_visits_outliers.sort_values('Doc_visits')
```

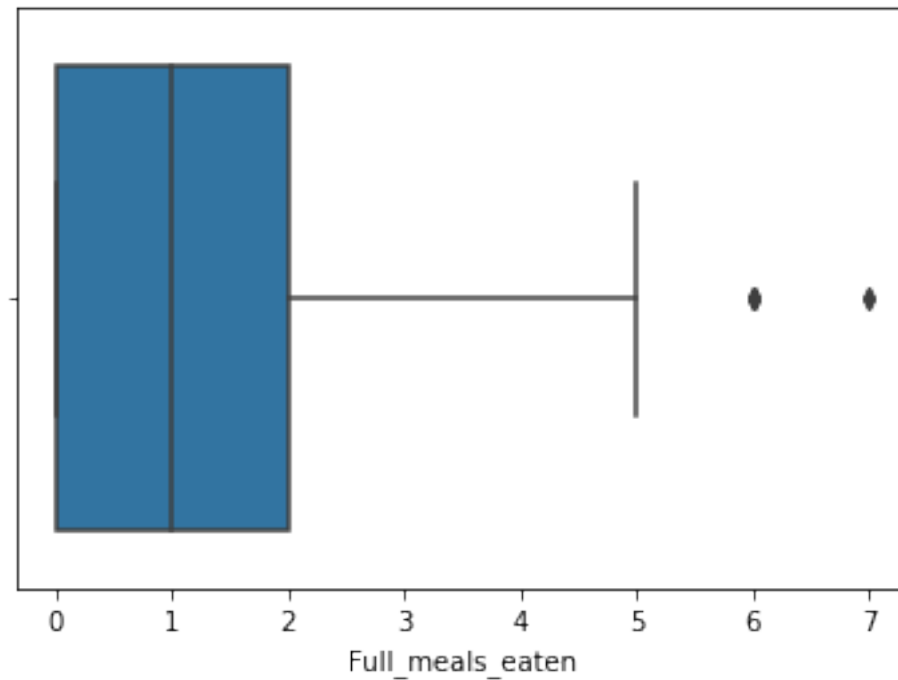
```
[70]:      Doc_visits  Doc_visits_z
Case_order
5646             1    -3.836921
5757             1    -3.836921
6018             1    -3.836921
6499             1    -3.836921
6943             1    -3.836921
7144             1    -3.836921
963              9     3.813587
2767             9     3.813587
```

```
[71]: doc_visits_outliers.sort_values('Doc_visits').value_counts()
```

```
[71]: Doc_visits  Doc_visits_z
1          -3.836921         6
9           3.813587         2
dtype: int64
```

```
[72]: sns.boxplot(x=numeric_data['Full_meals_eaten'])
```

[72]: <AxesSubplot:xlabel='Full_meals_eaten'>



```
[73]: full_meals_eaten_outliers = numeric_data.loc[
      (numeric_data['Full_meals_eaten_z'] > 3) |
      →(numeric_data['Full_meals_eaten_z'] < -3),
      ['Full_meals_eaten', 'Full_meals_eaten_z']]
Add_outlier_column(df, full_meals_eaten_outliers, 'Full_meals_eaten')
full_meals_eaten_outliers.sort_values('Full_meals_eaten')
```

```
[73]:
```

	Full_meals_eaten	Full_meals_eaten_z
Case_order		
551	5	3.966603
9068	5	3.966603
8995	5	3.966603
8903	5	3.966603
8327	5	3.966603
6803	5	3.966603
6695	5	3.966603
6084	5	3.966603
6027	5	3.966603
5860	5	3.966603
5712	5	3.966603
5598	5	3.966603
9221	5	3.966603

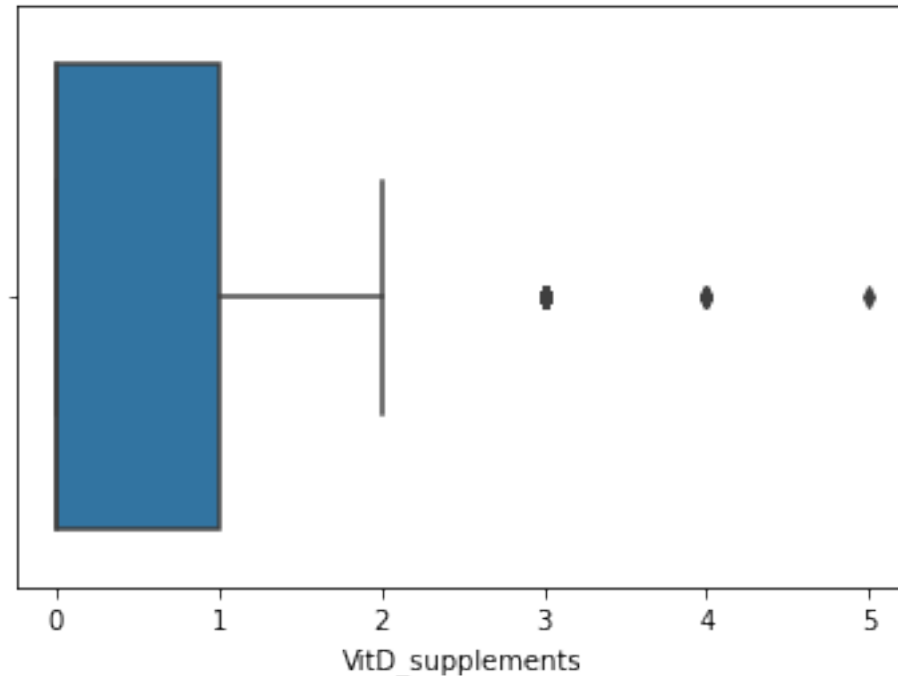
5368	5	3.966603
5544	5	3.966603
4346	5	3.966603
2920	5	3.966603
2878	5	3.966603
2747	5	3.966603
2653	5	3.966603
698	5	3.966603
2316	5	3.966603
4903	5	3.966603
1457	5	3.966603
1149	5	3.966603
2185	6	4.958602
1232	6	4.958602
9987	6	4.958602
7218	6	4.958602
6069	6	4.958602
8145	6	4.958602
959	7	5.950600
4710	7	5.950600

```
[74]: full_meals_eaten_outliers.sort_values('Full_meals_eaten').value_counts()
```

```
[74]: Full_meals_eaten  Full_meals_eaten_z
5                    3.966603            25
6                    4.958602             6
7                    5.950600             2
dtype: int64
```

```
[75]: sns.boxplot(x=numeric_data['VitD_supplements'])
```

```
[75]: <AxesSubplot:xlabel='VitD_supplements'>
```

```
[76]: vitD_supplements_outliers = numeric_data.loc[
        (numeric_data['VitD_supplements_z'] > 3) |
        (numeric_data['VitD_supplements_z'] < -3),
        ['VitD_supplements', 'VitD_supplements_z']]
Add_outlier_column(df, vitD_supplements_outliers, 'VitD_Supplements')
vitD_supplements_outliers.sort_values('VitD_supplements')
```

```
[76]:      VitD_supplements  VitD_supplements_z
Case_order
63                3      4.138759
5000               3      4.138759
5045               3      4.138759
5217               3      4.138759
5352               3      4.138759
...
1343               4      5.729917
9092               4      5.729917
7181               4      5.729917
2534               4      5.729917
3132               5      7.321074

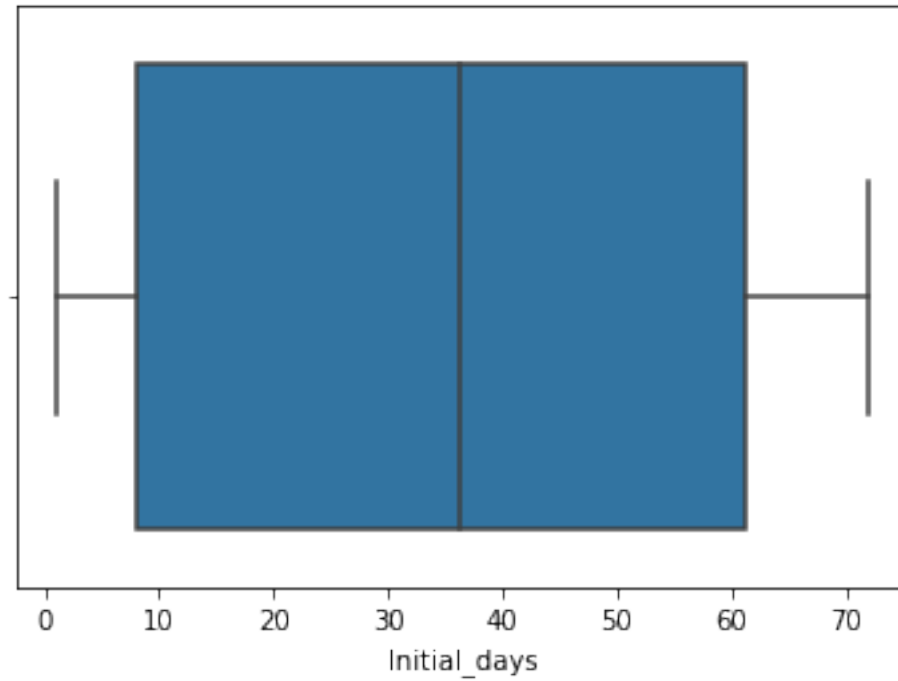
[70 rows x 2 columns]
```

```
[77]: vitD_supplements_outliers.sort_values('VitD_supplements').value_counts()
```

```
[77]: VitD_supplements  VitD_supplements_z
      3                4.138759          64
      4                5.729917           5
      5                7.321074           1
      dtype: int64
```

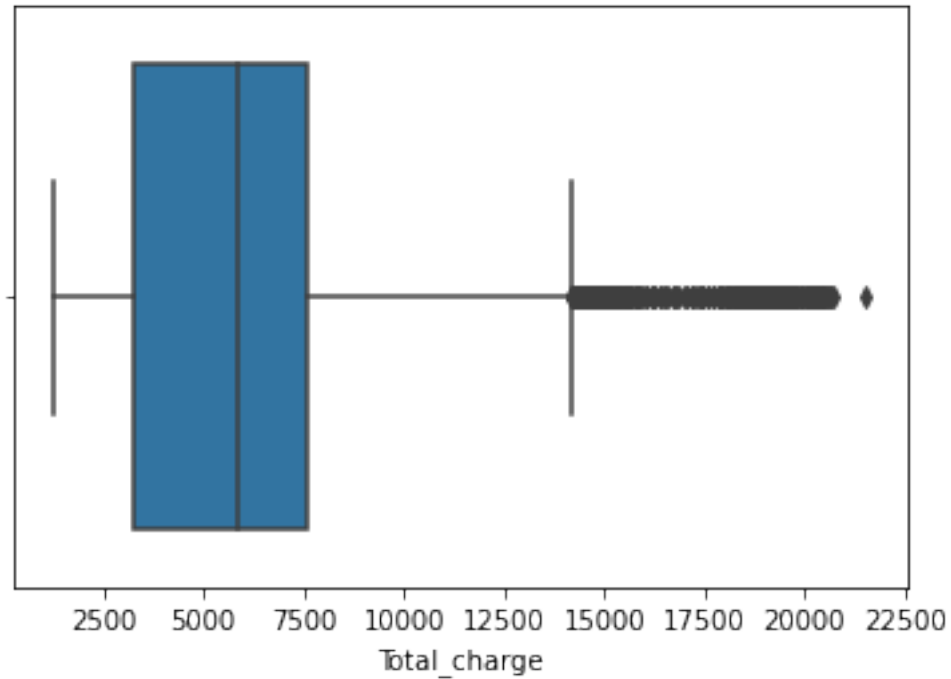
```
[78]: sns.boxplot(x=numeric_data['Initial_days'])
```

```
[78]: <AxesSubplot:xlabel='Initial_days'>
```



```
[79]: sns.boxplot(x=numeric_data['Total_charge'])
```

```
[79]: <AxesSubplot:xlabel='Total_charge'>
```



```
[80]: total_charge_outliers = numeric_data.loc[(numeric_data['Total_charge_z'] > 3) |
↳ (numeric_data['Total_charge_z'] < -3),
      ['Total_charge', 'Total_charge_z']]
Add_outlier_column(df, total_charge_outliers, 'Total_Charge')
total_charge_outliers.sort_values('Total_charge')
```

```
[80]:
```

	Total_charge	Total_charge_z
Case_order		
528	16053.46	3.008810
3351	16057.31	3.009950
1848	16153.99	3.038575
3000	16173.62	3.044388
1964	16194.01	3.050425
...
9160	20562.04	4.343740
5454	20632.44	4.364585
5245	20647.39	4.369011
9006	20673.97	4.376881
8801	21524.22	4.628629

[276 rows x 2 columns]

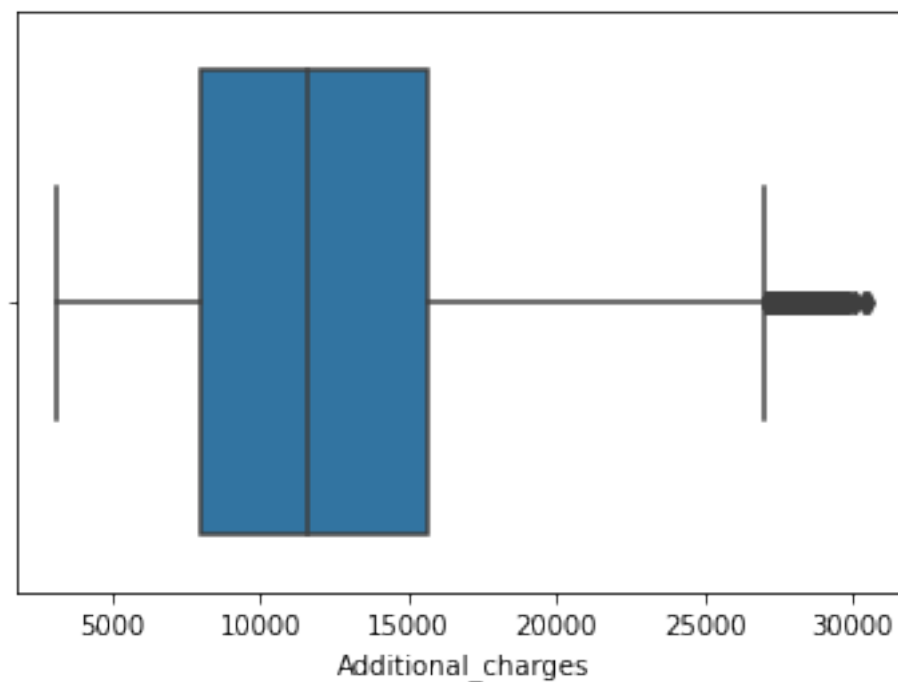
```
[81]: total_charge_outliers.sort_values('Total_charge').value_counts()
```

```
[81]: Total_charge  Total_charge_z
      16053.46      3.008810      1
      19367.21      3.989967      1
      19409.18      4.002394      1
      19404.99      4.001153      1
      19403.19      4.000620      1
      ..
      18550.12      3.748038      1
      18557.70      3.750282      1
      18564.13      3.752186      1
      18575.97      3.755691      1
      21524.22      4.628629      1
      Length: 276, dtype: int64
```

iqr used because z score did not accurately capture outliers due to data distribution

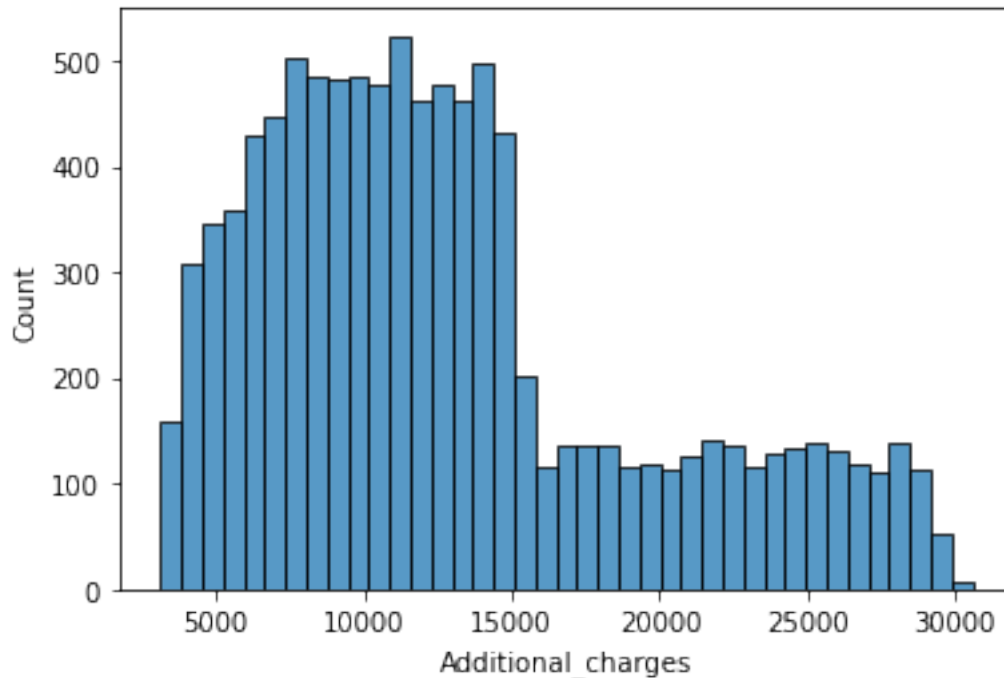
```
[82]: sns.boxplot(x=numeric_data['Additional_charges'])
```

```
[82]: <AxesSubplot:xlabel='Additional_charges'>
```



```
[83]: sns.histplot(x=numeric_data['Additional_charges'])
```

```
[83]: <AxesSubplot:xlabel='Additional_charges', ylabel='Count'>
```



```
[84]: iqr_a_charges = stats.iqr(numeric_data['Additional_charges'])
q1_a_charges = numeric_data['Additional_charges'].quantile(0.25)
q3_a_charges = numeric_data['Additional_charges'].quantile(0.75)
additional_charges_outliers = numeric_data.loc[
    (numeric_data['Additional_charges'] > (q3_a_charges * 1.5 + iqr_a_charges)) |
    (numeric_data['Additional_charges'] < (q1_a_charges * 1.5 - iqr_a_charges)),
    ['Additional_charges']]
Add_outlier_column(df, additional_charges_outliers, 'Additional_charges')
additional_charges_outliers.sort_values('Additional_charges')
```

```
[84]:
```

Case_order	Additional_charges
6452	3125.70
2415	3132.26
1478	3132.26
4232	3139.05
3515	3139.05
...	...
6465	4327.02
5288	4327.02
7279	4332.13
4650	4334.52
4539	4337.80

```
[383 rows x 1 columns]
```

```
[85]: additional_charges_outliers.sort_values('Additional_charges').value_counts()
```

```
[85]: Additional_charges
3241.34      4
3585.74      3
3883.66      3
4228.07      3
4129.06      3
..
3771.31      1
3767.15      1
3764.25      1
3760.09      1
4337.80      1
Length: 334, dtype: int64
```

##Re-expression of catagorical variables: –categorical columns that can only be yes or no will be converted to 1, or 0. –categorical columns that can be expressed ordinally will have a numerical column added with their ordinal value, in the format __numeric. –categorical columns that do not fit either of the prior categories will not be altered and will be retained for use in data analysis as is.

```
[86]: df.columns
```

```
[86]: Index(['City', 'State', 'County', 'Zip', 'Population', 'Area', 'Timezone',
        'Job', 'Children', 'Age', 'Education', 'Employment', 'Income',
        'Marriage_status', 'Gender', 'Readmitted', 'VitD_levels', 'Doc_visits',
        'Full_meals_eaten', 'VitD_supplements', 'Habitual_soft_drink_use',
        'Initial_admin', 'High_blood_pressure', 'Stroke', 'Complication_risk',
        'Overweight', 'Arthritis', 'Diabetes', 'Hyperlipidemia', 'Back_pain',
        'Anxiety', 'Allergic_rhinitis', 'Reflux_esophagitis', 'Asthma',
        'Primary_service_recived', 'Initial_days', 'Total_charge',
        'Additional_charges', 'Survey_timely_admission',
        'Survey_timely_treatment', 'Survey_timely_visits', 'Survey_reliability',
        'Survey_options', 'Survey_hours', 'Survey_courtesy',
        'Survey_active_listening', 'Population_outliers', 'Children_outliers',
        'Income_outliers', 'VitD_levels_outliers', 'Doc_visits_outliers',
        'Full_meals_eaten_outliers', 'VitD_Supplements_outliers',
        'Total_Charge_outliers', 'Additional_charges_outliers'],
        dtype='object')
```

```
[87]: df.loc[:, ['Readmitted', 'Habitual_soft_drink_use', 'High_blood_pressure',
        ↪ 'Stroke', 'Arthritis', 'Diabetes',
        ↪ 'Hyperlipidemia', 'Back_pain', 'Allergic_rhinitis',
        ↪ 'Reflux_esophagitis', 'Asthma']].replace({'Yes': 1, 'No': 0})
```

```
[87]:
```

	Readmitted	Habitual_soft_drink_use	High_blood_pressure	Stroke	\
Case_order					
1	0	0	1	0	
2	0	0	1	0	
3	0	0	1	0	
4	0	0	0	1	
5	0	1	0	0	
...	
9996	0	0	1	0	
9997	1	0	1	0	
9998	1	1	1	0	
9999	1	0	0	0	
10000	1	0	0	0	

	Arthritis	Diabetes	Hyperlipidemia	Back_pain	Allergic_rhinitis	\
Case_order						
1	1	1	0	1		1
2	0	0	0	0		0
3	0	1	0	0		0
4	1	0	0	0		0
5	0	0	1	0		1
...	
9996	0	0	0	0		0
9997	1	1	0	0		0
9998	0	0	0	0		1
9999	0	0	0	1		0
10000	1	0	1	0		1

	Reflux_esophagitis	Asthma
Case_order		
1	0	1
2	1	0
3	0	0
4	1	1
5	0	0
...
9996	1	0
9997	0	1
9998	0	0
9999	0	0
10000	0	0

[10000 rows x 11 columns]

```
[88]: education_dict = {'Some College, Less than 1 Year': 5,
                        'Some College, 1 or More Years, No Degree': 6,
                        'GED or Alternative Credential': 3, 'Regular High School Diploma': 4,
```

```

    "Bachelor's Degree": 9, "Master's Degree": 10,
    'Nursery School to 8th Grade': 1,
    '9th Grade to 12th Grade, No Diploma': 2, 'Doctorate Degree': 11,
    "Associate's Degree": 8, 'Professional School Degree': 7,
    'No Schooling Completed': 0}
df['Education_numeric'] = df['Education'].replace(education_dict)
df['Education_numeric'].value_counts()

```

```

[88]: 4      2444
      9      1724
      6      1484
      2       832
      8       797
     10       701
      5       642
      1       552
      3       389
      7       208
      0       133
     11        94
      Name: Education_numeric, dtype: int64

```

```

[89]: complication_dict = {'Low': 0, 'Medium': 1, 'High': 2}
df['Complication_risk_numeric'] = df['Complication_risk'].
    ↪replace(education_dict)
df['Complication_risk_numeric'].value_counts()

```

```

[89]: Medium      4517
      High       3358
      Low        2125
      Name: Complication_risk_numeric, dtype: int64

```

Principal component Analysis

```
[ ]:
```