

Classification Models to Identify Water Pump Functionality in Tanzania

Overview

This project consists of classification models to identify water pump functionality in Tanzania. Limited access to safe water is a major health risk. These models are potential tools for government agencies or non-governmental organizations to identify areas with limited access to clean water. Through an iterative modeling process, we produced a model that predicted pump functionality with 85% accuracy. By identifying non-functional water pumps, organizations can divert resources to areas in need of assistance and improve water access and health in Tanzania.

Business Understanding

Millions of people in Tanzania lack access to safe water. This results in paying high prices for water from vendors or collecting water from unsafe natural sources. In order to combat this problem, resources must be allocated to fix non-functioning water distribution points. Age is an important metric in predicting the condition of distribution points. Older pumps and engine systems are more likely to fail than newer ones. In many cases age related data is not available. The goal of this analysis is to build the model that can predict the condition of waterpoints based on their other features such as regional factors, installer, type of pump, population and others

Data Understanding

The data was sourced from the Taarifa waterpoint dashboard, which aggregates data from the Tanzania Ministry of Water. The information collected was recorded by GeoData Consultants Ltd. There are 59,400 rows and 40 columns in the "water_well_train_data.csv".

Our target data is stored in "water_well_train_labels.csv". There are 59,400 rows and 2 columns in this csv file. The two columns in this csv file are 'id' and 'status_group'. The 'id' column aligns with the 'id' column in the "water_well_train_data.csv" file. Our target column is 'status_group' which consists of three values describing the status of a water pump: "functional", "functional needs repair", and "non-functional".

Data Cleaning and EDA

In [1]:

```
#import statements
import pandas as pd
import numpy as np

#data visualization
import matplotlib.pyplot as plt
import seaborn as sns

#sci-kit learn
import sklearn
from sklearn.model_selection import train_test_split, cross_validate, cross_val_score, GridSearchCV
from sklearn.preprocessing import FunctionTransformer, OneHotEncoder, StandardScaler
from sklearn.metrics import accuracy_score, plot_confusion_matrix
from sklearn.dummy import DummyClassifier

from sklearn.neighbors import KNeighborsClassifier, NearestNeighbors
from sklearn.tree import DecisionTreeClassifier
```

```
from sklearn.ensemble import RandomForestClassifier

from sklearn.pipeline import Pipeline
from sklearn.compose import ColumnTransformer
```

In [2]:

```
#import train data
#DO NOT LOOK AT TEST DATA UNTIL VALIDATION
df_train = pd.read_csv('./data/water_well_train_data.csv')
```

In [3]:

```
df_train.head()
```

Out[3]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	pay
0	69572	6000.0	2011-03-14	Roman	1390	Roman	34.938093	-9.856322	none	0	...	
1	8776	0.0	2013-03-06	Grumeti	1399	GRUMETI	34.698766	-2.147466	Zahanati	0	...	
2	34310	25.0	2013-02-25	Lottery Club	686	World vision	37.460664	-3.821329	Kwa Mahundi	0	...	
3	67743	0.0	2013-01-28	Unicef	263	UNICEF	38.486161	11.155298	Zahanati Ya Nanyumbu	0	...	
4	19728	0.0	2011-07-13	Action In A	0	Artisan	31.130847	-1.825359	Shuleni	0	...	

5 rows x 40 columns



In [4]:

```
# info of train data
df_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 59400 entries, 0 to 59399
Data columns (total 40 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   id                                     59400 non-null  int64
1   amount_tsh                           59400 non-null  float64
2   date_recorded                         59400 non-null  object
3   funder                                55765 non-null  object
4   gps_height                            59400 non-null  int64
5   installer                             55745 non-null  object
6   longitude                             59400 non-null  float64
7   latitude                              59400 non-null  float64
8   wpt_name                              59400 non-null  object
9   num_private                           59400 non-null  int64
10  basin                                 59400 non-null  object
11  subvillage                            59029 non-null  object
12  region                                 59400 non-null  object
13  region_code                           59400 non-null  int64
14  district_code                         59400 non-null  int64
15  lga                                    59400 non-null  object
16  ward                                  59400 non-null  object
17  population                             59400 non-null  int64
18  public_meeting                        56066 non-null  object
19  recorded_by                           59400 non-null  object
20  scheme_management                     55523 non-null  object
21  scheme_name                           31234 non-null  object
22  permit                                56344 non-null  object
23  construction_year                     59400 non-null  int64
24  ...
```

```
24  extraction_type           59400 non-null object
25  extraction_type_group    59400 non-null object
26  extraction_type_class    59400 non-null object
27  management               59400 non-null object
28  management_group         59400 non-null object
29  payment                   59400 non-null object
30  payment_type              59400 non-null object
31  water_quality             59400 non-null object
32  quality_group             59400 non-null object
33  quantity                  59400 non-null object
34  quantity_group            59400 non-null object
35  source                    59400 non-null object
36  source_type               59400 non-null object
37  source_class              59400 non-null object
38  waterpoint_type           59400 non-null object
39  waterpoint_type_group     59400 non-null object
dtypes: float64(3), int64(7), object(30)
memory usage: 18.1+ MB
```

The data contains 59,400 rows and 40 columns.

Our target column is stored in a separate csv file.

In [5]:

```
#import target information
df_label = pd.read_csv('./data/water_well_train_labels.csv')
```

In [6]:

```
df_label.shape
```

Out[6]:

```
(59400, 2)
```

In [7]:

```
df_label.head()
```

Out[7]:

	id	status_group
0	69572	functional
1	8776	functional
2	34310	functional
3	67743	non functional
4	19728	functional

In [8]:

```
df_label.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 59400 entries, 0 to 59399
Data columns (total 2 columns):
#   Column          Non-Null Count  Dtype
---  -
0   id               59400 non-null  int64
1   status_group     59400 non-null  object
dtypes: int64(1), object(1)
memory usage: 928.2+ KB
```

In [9]:

```
df_label['status_group'].value_counts()
```

Out[9]:

Out[9]:

```
functional          32259
non functional      22824
functional needs repair    4317
Name: status_group, dtype: int64
```

In [10]:

```
df_label['status_group'].value_counts(normalize = True)
```

Out[10]:

```
functional          0.543081
non functional      0.384242
functional needs repair    0.072677
Name: status_group, dtype: float64
```

There are three target classifications: functional (54%), non-functional (38%), and function needs repair (7%). We combined the 'status_group' dataframe with the train_data dataframe.

In [11]:

```
#combine train and label dataframes prior to cleaning to address any dropped rows
df = df_train.join(other = df_label, rsuffix = '_label')
```

In [12]:

```
df.head()
```

Out[12]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
0	69572	6000.0	2011-03-14	Roman	1390	Roman	34.938093	-9.856322	none	0	...	
1	8776	0.0	2013-03-06	Grumeti	1399	GRUMETI	34.698766	-2.147466	Zahanati	0	...	
2	34310	25.0	2013-02-25	Lottery Club	686	World vision	37.460664	-3.821329	Kwa Mahundi	0	...	
3	67743	0.0	2013-01-28	Unicef	263	UNICEF	38.486161	11.155298	Zahanati Ya Nanyumbu	0	...	
4	19728	0.0	2011-07-13	Action In A	0	Artisan	31.130847	-1.825359	Shuleni	0	...	

5 rows x 42 columns



A quick confirmation to see that the columns in the dataframe are properly aligned. The id values from each dataframe should match.

In [13]:

```
#check that id columns align
df[df['id'] != df['id_label']]
```

Out[13]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	quality_group
--	----	------------	---------------	--------	------------	-----------	-----------	----------	----------	-------------	-----	---------------

0 rows x 42 columns



Now that our features and target were in the same dataframe, we could begin exploring and cleaning the data.

In [14]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 59400 entries, 0 to 59399
Data columns (total 42 columns):
#   Column                               Non-Null Count  Dtype
---  -
0   id                                    59400 non-null  int64
1   amount_tsh                           59400 non-null  float64
2   date_recorded                         59400 non-null  object
3   funder                                55765 non-null  object
4   gps_height                            59400 non-null  int64
5   installer                             55745 non-null  object
6   longitude                             59400 non-null  float64
7   latitude                              59400 non-null  float64
8   wpt_name                              59400 non-null  object
9   num_private                           59400 non-null  int64
10  basin                                 59400 non-null  object
11  subvillage                            59029 non-null  object
12  region                                59400 non-null  object
13  region_code                           59400 non-null  int64
14  district_code                         59400 non-null  int64
15  lga                                    59400 non-null  object
16  ward                                  59400 non-null  object
17  population                             59400 non-null  int64
18  public_meeting                        56066 non-null  object
19  recorded_by                           59400 non-null  object
20  scheme_management                     55523 non-null  object
21  scheme_name                           31234 non-null  object
22  permit                                56344 non-null  object
23  construction_year                     59400 non-null  int64
24  extraction_type                       59400 non-null  object
25  extraction_type_group                  59400 non-null  object
26  extraction_type_class                  59400 non-null  object
27  management                             59400 non-null  object
28  management_group                       59400 non-null  object
29  payment                                59400 non-null  object
30  payment_type                           59400 non-null  object
31  water_quality                          59400 non-null  object
32  quality_group                          59400 non-null  object
33  quantity                               59400 non-null  object
34  quantity_group                         59400 non-null  object
35  source                                 59400 non-null  object
36  source_type                            59400 non-null  object
37  source_class                           59400 non-null  object
38  waterpoint_type                        59400 non-null  object
39  waterpoint_type_group                  59400 non-null  object
40  id_label                               59400 non-null  int64
41  status_group                           59400 non-null  object
dtypes: float64(3), int64(8), object(31)
memory usage: 19.0+ MB
```

We have 42 columns: the majority of them were object type with a few numerical.

Columns with nulls:

- **funder**
- **installer**
- **subvillage**
- **public_meeting**
- **scheme_management**
- **scheme_name**
- **permit**

In [15]:

```
df.describe()
```

Out[15]:

	id	amount_tsh	gps_height	longitude	latitude	num_private	region_code	district_code	
count	59400.000000	59400.000000	59400.000000	59400.000000	5.940000e+04	59400.000000	59400.000000	59400.000000	59
mean	37115.131768	317.650385	668.297239	34.077427	5.706033e+00	0.474141	15.297003	5.629747	
std	21453.128371	2997.574558	693.116350	6.567432	2.946019e+00	12.236230	17.587406	9.633649	
min	0.000000	0.000000	-90.000000	0.000000	1.164944e+01	0.000000	1.000000	0.000000	
25%	18519.750000	0.000000	0.000000	33.090347	8.540621e+00	0.000000	5.000000	2.000000	
50%	37061.500000	0.000000	369.000000	34.908743	5.021597e+00	0.000000	12.000000	3.000000	
75%	55656.500000	20.000000	1319.250000	37.178387	3.326156e+00	0.000000	17.000000	5.000000	
max	74247.000000	350000.000000	2770.000000	40.345193	-2.000000e-08	1776.000000	99.000000	80.000000	30

Dealing with Nulls

We looked at every column with null values to determine how to address those missing values.

funder

In [16]:

```
#who funded the well
print(df['funder'].isna().sum())
df['funder'].value_counts()
```

3635

Out[16]:

Government Of Tanzania	9084
Danida	3114
Hesawa	2202
Rwssp	1374
World Bank	1349
...	
Pentekoste	1
Usambala Sister	1
Rumaki	1
Friedkin Conservation Fund	1
Rashid Seng'ombe	1
Name: funder, Length: 1897, dtype: int64	

There were 3,635 nulls in 'funder'.

In [17]:

```
unique_funder = list(df['funder'].unique())
print(len(unique_funder))
unique_funder
```

1898

Out[17]:

['Roman',
'Grumeti',
'Mottawa Club']

'Lottery Club',
'Unicef',
'Action In A',
'Mkinga Distric Coun',
'Dwsp',
'Rwssp',
'Wateraid',
'Isingiro Ho',
'Private',
'Danida',
'World Vision',
'Lawatefuka Water Supply',
'Biore',
'Rudep',
'Hesawa',
'Twe',
'Isf',
'African Development Bank',
'Government Of Tanzania',
'Sobodo',
'Water',
'Private Individual',
'Undp',
nan,
'Not Known',
'Kirde',
'Cefa',
'Ces(gmbh)',
'European Union',
'Lga',
'District Council',
'Muwsa',
'Dwe/norad',
'Kkkt_makwale',
'Sawaka',
'Ces (gmbh)',
'Olgilai Village Community',
'Kkkt',
'Roman Catholic',
'Norad',
'Adra',
'Sema',
'Piusi',
'Dwe',
'Rc Church',
'Swisland/ Mount Meru Flowers',
'Ifad',
'Swedish',
'Idc',
'He',
'Isf/tacare',
'Jica',
'Mzee Sh',
'Aict',
'Tcrs',
'Kiuma',
'Germany Republi',
'Netherlands',
'Ruthe',
'Tulawaka Gold Mine',
'Nethalan',
'Tasaf',
'Concern World Wide',
'Wfp',
'Lips',
'Sida',
'World Bank',
'Tanza',
'0',
'Sw',
'Shipo',
'Fini Water',
'Zambia'

'Kaniisa',
'Oxfarm',
'Village Council',
'Hesawz',
'Shanta',
'Fpct',
'Wvt',
'Dhv',
'Ir',
'Oikos E.Afrika',
'Anglican Church',
'Peters',
'Donor',
'Secondary Schoo',
'Amref',
'Ministry Of Water',
'Adb',
'Jbg',
'Dadis',
'International Aid Services',
'Germany',
'Kibaha Town Council',
'Dsdp',
'Dfid',
'Rural Water Supply And Sanitat',
'Af',
'Wananchi',
'Fw',
'No',
'Dct',
'Africare',
'Norad /government',
'British Colonial Government',
'Co',
'Ridep',
'Tassaf',
'Hans',
'Socie',
'Finw',
'Fin Water',
'Oxfam',
'Plan International',
'African Muslim Agency',
'Go',
'Cdtf',
'Shawasa',
'Un',
'Awf',
'Commu',
'Community',
'Save The Rain Usa',
'Kibara Foundation',
'Tlc',
'Rc Churc',
'Plan Int',
'W.B',
'Lvvia',
'Songea District Council',
'Hifab',
'Rc Ch',
'Makonde Water Population',
'Snv',
'Government/ Community',
'National Rural',
'Is',
'Giz',
'Cspd',
'Medicine',
'Wsdp',
'Unice/ Cspd',
'Finn Water',
'Kamama',
'W.B',

'Villagers',
'Ereto',
'Abasia',
'Unhcr',
'Ebaha',
'Kuwait',
'Magadini-makiwaru Water',
'Mh An',
'Kaemp',
'African Relie',
'Rcchurch/cefa',
'Norad/ Kidep',
'Private Owned',
'Tardo',
'Insututional',
'Sabemo',
'Missi',
'Dmdd',
'Dhv\\norp',
'Mission',
'Ru',
'Halmashauri Ya Wilaya Sikonge',
'Japan',
'Simone',
'Ki',
'Peace Cope',
'Finland',
'Marafip',
'Ta',
'Su-ki Jang',
'Tado',
'Tanzania',
'Il',
'Bank',
'Ded',
'Irc',
'Sabodo',
'Soda',
'I.E.C',
'Drdp Ngo',
'Lwi',
'Maxavella',
'Ics',
'African',
'Bilila',
'Tpp',
'Cipro/government',
'Tabora Municipal Council',
'Salim Ahmed Salim',
'Eu/acra',
'Kadres Ngo',
'Regional Water Engineer Arusha',
'Quickwi',
'Dhv Moro',
'Hewasa',
'Tasaf And Lga',
'Jaica',
'Village Res',
'Kkkt-dioces Ya Pare',
'Aic',
'Solidarm',
'Christan Outrich',
'Kanisa La Menonite',
'Islamic',
'Rc',
'Killflora',
'Bread For The Wor',
'Wua',
'Mac',
'Caltaz Kahama',
'Mianz',
'Dw',
'Mabamashini'

'Makapuchini',
'Tasaf Ii',
'Omar Ally',
'Md',
'Mitema',
'Ham',
'Quwkwin',
'Do',
'Dh',
'Bokera W',
'Bulyahunlu Gold Mine',
'Mbiuwasa',
'The Isla',
'Rotary Club',
'Muslims',
'Care International',
'Kimkuma',
'Tanesco',
'Mbozi District Council',
'Dasip',
'Tltc',
'Sdg',
'Hsw',
'Mwaya Mn',
'Resolute Mining',
'Tz Japan',
'Roman Cathoric Same',
'Concern',
'Caritas',
'Conce',
'Huches',
'Wamarekani',
'Devon Aid Korogwe',
'Kiliwater',
'Lamp',
'Bsf',
'Mem',
'Jeica',
'Father Bonifasi',
'Bgm',
'Lcgd',
'Karadea Ngo',
'Msf/tacare',
'Fathe',
'Unice',
'Mdc',
'Dasp',
'Songea Municipal Council',
'Tasae',
'Water User As',
'Msikiti',
'Cct',
'Islamic Found',
'Tgrs',
'Unicef/ Csp',
'Jimbo Fund',
'Tlc/john Majala',
'Magoma Adp',
'Vwc',
'Pidp',
'Japan Government',
'Kata',
'De',
'Acra',
'Gtz',
'Isf/government',
'Kuwasa',
'China Government',
'Taboma',
'P',
'Kingupira S',
'Churc',
'W-1-1-1-1-1'

'Walokole',
'Mkinga Distric Cou',
'Cafod',
'Hw/rc',
'Sumbawanga Munici',
'Tacare',
'Urt',
'Camavita',
'Member Of Parliament',
'Dmmd',
'Aqua Blues Angels',
'Water Aid /sema',
'Kirdep',
'Cc Motor Day 2010',
'Kilwater',
'Ndrdp',
'Hez',
'Nethe',
'Denat',
'Kibo Brewaries',
'Arab Community',
'Elct',
'Adp',
'Priva',
'Holland',
'Rc Church/centr',
'Cocen',
'Wfp/tnt',
'Lench Taramai',
'Ncaa',
'Mzee Don',
'World Vision/ Kkkt',
'Finwater',
'Kuamu',
'Dwssp',
'Musilim Agency',
'Ukiligu',
'Wamakapuchini',
'Mbunge',
'The Desk And Chair Foundat',
'Duwas',
'Diwani',
'Kkkt Church',
'Ea',
'Halmashauri Ya Manispa Tabora',
'Finidagermantanzania Govt',
'Bahewasa',
'Jika',
'Asb',
'Qwiqwi',
'Pmo',
'Tuwasa',
'Irish Ai',
'Mdrdp',
'Jeshi La Wokovu',
'Government /tassaf',
'Mboma',
'People From Japan',
'Kilindi District Co',
'Shamte Said',
'Auwasa',
'Kidp',
'Tridep',
'St',
'Wd And Id',
'Serikali',
'Kanisa Katoliki',
'Po',
'Ga',
'Cocern',
'Finida German Tanzania Govt',
'National Rural And Hfa',
'vz'

'A',
'Idara Ya Maji',
'Moslem Foundation',
'Swiss If',
'Miziriol',
'Yasini Selemani',
'Dbspe',
'H',
'A/co Germany',
'Oikos E.Africa/european Union',
'Hydom Luthelani',
'Ilct',
'Peter Tesha',
'Ms',
'Mzungu Paul',
'Caltas',
'Red Cross',
'Losaa-kia Water Supply',
'Tassaf I',
'Kanisa Katoliki Lolovoni',
'Finland Government',
'Gaica',
'Institution',
'Tcrs.Tlc',
'Magereza',
'Loliondo Parish',
'Diocese Of Geita',
'Total Landcare',
'U.S.A',
'Tdft',
'Parastatal',
'Rished',
'Dwt',
'The People Of Japan',
'Kcu',
'Abd',
'Village Government',
'Msabi',
'Vc',
'Cmsr',
'Konoike',
'Roman Catholic Rulenge Diocese',
'Bened',
'Shule',
'W',
'Partage',
'Inkinda',
'Robert Loyal',
'Africa Amini Alama',
'Imf',
'L',
'Moroil',
'Sekei Village Community',
'Us Embassy',
'Missionaries',
'Tcrs /government',
'Desk And Chair Foundation',
'Ms-danish',
'Wsdp & Sdg',
'Roman Cathoric-same',
'Cefa-njombe',
'Aar',
'Village Govt',
'Farm Africa',
'Mheza Distric Counc',
'Chamavita',
'Mileniam Project',
'Undp/ilo',
'Dads',
'Institutional',
'Sowasa',
'Ccpg',
'm--s--'

'Tasalu',
'Government/ World Bank',
'Luthe',
'Wirara Ya Maji',
'Mzee Mkungata',
'Rada',
'Twesa',
'Plan Internatio',
'Solidame',
'Rwsso',
'Williamson Diamond Ltd',
'Tag',
'Dar Al Ber',
'Watu Wa Ujerumani',
'Dwe/bamboo Projec',
'Danida /government',
'Semaki K',
'Arabs Community',
'Water Aid/sema',
'District Rural Project',
'Gen',
'Redep',
'Kiwanda Cha Samaki',
'Singida Yetu',
'Rwsp',
'Moravian',
'Sema S',
'Cbhi',
'Tcrs /care',
'Makonde',
'Millenium',
'Swisland/mount Meru Flowers',
'Kigoma Municipal',
'Kinapa',
'People Of Japan',
'Kijij',
'Wfp/tnt/usaaid',
'Tanapa',
'Efg',
'Local',
'Kyariga',
'Tanzakesho',
'Roman Cathoric -kilomeni',
'World Vision/adra',
'Mbozi Secondary School',
'Tasaf/dmdd',
'Mws',
'Shekhe',
'Pataji',
'Tahea',
'Kalta',
'Pentecosta Church',
'Sekondari',
'Kyela Council',
'Kalitasi',
'Quick Wins',
'Lowasa',
'Hotels And Loggs Tz Ltd',
'Cobashec',
'Orphanage',
'Adf',
'Wwf',
'Idydc',
'Cper',
'School',
'Ilo',
'Olumuro',
'Villaers',
'Tlc/thimotheo Masunga',
'Dak',
'Kidep',
'Ubalози WaMarekani',
'Tale ...'

'Dmk Anglican',
'Franc',
'Ka',
'Mgm',
'Aimgold',
'Mzee Omari',
'Petro Patrice',
'Camartec',
'Loliondo Secondary',
'Islamic Agency Tanzania',
'Tanz Egypt Technical Cooper',
'Safari Roya',
'Koica',
'Rdc',
'Total Land Care',
'Pad',
'Msf',
'Mamad',
'Padep',
'One Un',
'Fabia',
'Lake Tanganyika',
'Italy',
'Solar Villa',
'Roman Church',
'Singasinga',
'Rc/mission',
'In',
'Adp Mombo',
'Pci',
'Norad/ Tassaf Ii',
'I Wash',
'Bs',
'Kambi Migoko',
'Ai',
'Sauwasa',
'Icdp',
'Rotte',
'Dhv/gove',
'Kmcl',
'Ccps',
'Si',
'Rundu Man',
'Serikari',
'Undp/aict',
'Hdv',
'Halmashauri',
'Concern /govern',
'Quick Win Project /council',
'Mh Kapuya',
'Halmashauri Ya Wilaya',
'Baric',
'Cpro',
'Getekwe',
'Gain',
'Wahidi',
'Asdp',
'Kadp',
'Aco/germany',
'Majengo Prima',
'Hortanzia',
'Quick',
'Hasnan Murig (mbunge)',
'Ikeuchi Towels Japan',
'Halmashauli',
'Acord',
'Menon',
'Wate Aid/sema',
'Dwe/ubalozi Wa Marekani',
'Vifafi',
'Cdg',
'Kwasenenge Group',
'Dad /magan'

'Dea/rwssp',
'Oldonyolengai',
'None',
'Village Community',
'Minjingu',
'El',
'D',
'Songas',
'Mi',
'Action Aid',
'Tanroad',
'Lake Tanganyika Basin',
'Pwc',
'Teonas Wambura',
'Mgaya Masese',
'Stantons',
'Sao H',
'Ukida',
'Taasaf',
'Mwita Kichere',
'Lwf',
'Mosque',
'Peter Ngereka',
'Svn',
'Investor',
'Ju-sarang Church' And Bugango",
'Lgcdg',
'Action Contre La Faim',
'Kwamdulu Estate',
'Quick Wins Scheme',
'Cpps',
'Belgian Government',
'Cmcrr',
'Care Int',
'Mavuno Ngo',
'Niger',
'Mwanza',
'Zaburi And Neig',
'Women For Partnership',
'Artisan',
'Sisa',
'Cdcg',
'Ndm',
'Secondary',
'Da Unoperaio Siciliano',
'Town Council',
'Lions Club',
'Lutheran Church',
'Shirika La Kinamama Na Watot',
'Pangadeco',
'Uyoge',
'Canada',
'Frankfurt',
'Redet',
'Rural Water Department',
'Buptist',
'Unp/aict',
'Timothy Shindika',
'Village Office',
'Lotary Club',
'Hesaw',
'Malec',
'Kuji Foundation',
'Mamvua Kakungu',
'Rusumo Game Reserve',
'Mtuwasa And Community',
'W.D.&I.',
'Act Mara',
'Sda',
'Mzinga A',
'Vgovernment',
'Re',
'T - - - - -'

'Loccip',
'Sua',
'Brdp',
'Hamref',
'Happy Watoto Foundation',
'Gdp',
'Lgdcg',
'Jgb',
'Mfuko Wa Jimbo',
'Doddea',
'Maliasili',
'Roman Ca',
'Tcrst',
'Holla',
'African Development Foundation',
'Fptc - Pent',
'Makona',
'Oxfam Gb',
'African 2000 Network',
'Netherland',
'Tabraki',
'Balo',
'Dadp',
'Ikela Wa',
'Rotary I',
'Rwssp/wsdp',
'Christian Outrich',
'Cipro/care/tcrs',
'Italian',
'Kome Parish',
'Mwanga Town Water Authority',
'Jumanne Siabo',
'Hindu',
'Rural',
'H/w',
'Tanap',
'Roman Cathoric Church',
'Rombo Dalta',
'Ilwilo Community',
'Un/wfp',
'St Ph',
'Lwiji Italy',
'Livin',
'Cg',
'Hhesawa',
'Lwi & Central Government',
'Lc',
'Kkkt Leguruki',
'Tanzania Compasion',
'Louise Elucas Sala',
'Hiap',
'Cpps Mission',
'Matyenye',
'Dimon',
'Italy Government',
'Tag Church Ub',
'Aic Church',
'Wvc',
'Lgcbg',
'Tacri',
'Chai Wazir',
'Hasnein Muriij',
'Rural Water Supply And Sanita',
'Simba Lodge',
'Free Pentecoste Church Of Tanz',
'Summit For Water',
'Sanje Wa',
'Makundya',
'Uhai Wa Mama Na Mtoto',
'Ola',
'Ba As',
'Tredep',
'Mwanga Development'

'Nyanza Road',
'Cgc',
'Swidish',
'Kizenga',
'Hapa',
'Ramadhani Nyambizi',
'Denish',
'Mkuyu',
'Ras',
'Mwinjuma Mzee',
'Gachuma Ginery',
'Resolute',
'Morovian',
'Water Board',
'Kigoma Municipal Council',
'Mafwimbo',
'Pentecostal',
'Rocci Ross',
'Igolola Community',
'Pancrasi',
'S',
'Rdws',
'Said Omari',
'Ngiresi Village Community',
'Kilomber',
'Sharifa Athuman',
'Qwickwin',
'Mwita Muremi',
'Mbwana Omari',
'Tlc/samora',
'Mmem',
'Haydom Lutheran Hospital',
'Vicfish Ltd',
'Afroz Ismail',
'Sisal Estste Hale',
'Eu',
'Korea',
'Cvs Miss',
'Moradi',
'Living Water International',
'Kajima',
'Uaacc',
'Germany Misionary',
'Rips',
'France',
'Bukumbi',
'Rhobi',
'Kiwanda Cha Tangawizi',
'Ten Degree Hotel',
'Wssp',
'Meru Concrete',
'Gg',
'Wizara',
'Segera Estate',
'Hospital',
'Dmk',
'Siza Mayengo',
'Greec',
'Makli',
'Mp',
'Islam',
'Dassip',
'Rvemp',
'Adp Bungu',
'Thomasi Busigaye',
'Sijm',
'W.D & I.',
'British Tanza',
'Kkkt Ndrumangeni',
'Tag Church',
'Council',
'Usambala Sister',
'Wanda Mwanza Wanda Tanza'

'Hearts Helping Hands.Inc.',
'Idea',
'Filo',
'Qwekwin',
'Selous G',
'Pentecostal Hagana Sweeden',
'Ester Ndege',
'Oikos E .Africa/european Union',
'Nyabarongo Kegoro',
'Quik',
'Ringo',
'Kanisani',
'Wfp/usaaid/tnt',
'Village Council/ Haydom Luther',
'Fpct Church',
'Mzung',
'Kwikwiz',
'Kanisa La Mitume',
'Iom',
'Oda',
'Caltus',
'Gt',
'Malola',
'Water Project Mbawala Chini',
'Totoland Care',
'Nddp',
'Kmt',
'Anjuman E Seiffee',
'Nginila',
'Usa Embassy',
'Village',
'Pdi',
'T',
'Hery',
'Obc',
'Nyamongo Gold Mining',
'Women Fo Partnership',
'Sister Francis',
'Norani',
'Mahita',
'Kalebejo Parish',
'Aixos',
'Government',
'Wrssp',
'Ddp',
'Game Division',
'Rudep /dwe',
'Kashwas',
'Twende Pamoja',
'Gwitembe',
'Makori',
'Sangea District Council',
'Unicef/central',
'Africa 2000 Network/undp',
'Mmany Abdallah',
'Snv Ltd',
'Taes',
'Canada Aid',
'Senapa',
'Regwa Company Of Egypt',
'Water Se',
'Mamlaka Ya Maji Ngara',
'Wama',
'Prf',
'Church',
'Magadini Makiwaru Water',
'Kayempu Ltd',
'Trachoma',
'Seleman Rashid',
'Afriican Reli',
'Tassaf Ii',
'Samsoni',
'Quikwiz'

'Quick Wings',
'Ngos',
'Kurrp Ki',
'Cast',
'Rudep/norad',
'Kwa Mzee Waziri',
'Panone',
'Lawate Fuka Water Suppl',
'St Gasper',
'Wug And Ded',
'Pr',
'Mmg Gold Mine',
'Nordic',
'Mchukwi Hos',
'Dwst',
'Serikaru',
'African Realief Committe Of Ku',
'Fao',
'Scott',
'Mzungu',
'Vttp',
'Vi',
'Irish Government',
'Namungo Miners',
'Nassor Fehed',
'Dbfpe',
'Clause',
'Busoga Trust',
'Mzee Mabena',
'Br',
'Brad',
'Koico',
'Healt',
'Ro',
'Jeshi Lawokovu',
'Paffect Mwanaindi',
'Tansi',
'Craelius',
'Apm[africa Precious Metals Lt',
'Zao Water Spring X',
'Shinyanga Shallow Wells',
'Cipro/care',
'Vifaf',
'Mtc',
'Lungwe',
'Dhinu',
'Aic Kij',
'Mataro',
'Dagida',
'Redap',
'Nwssp',
'Lench',
'Wanakijiji',
'Nk',
'Nimrodi Mkono[mb]',
'Maro',
'Professor Ben Ohio University',
'Rafael Michael',
'Tdrs',
'Bra',
'Suwasa',
'Twig',
'Tanzania Egypt Technical Co Op',
'Lifetime',
'Comunedi Roma',
'Unhcr/danida',
'Bread Of The Worl',
'Lutheran',
'Tasf',
'Rc Cathoric',
'Halmashauri Wil',
'Mgaya',
'Quick Mission Mzee Waziri'

```

'Grail MISSION Kiseki Bar',
'Answer Muslim Grou',
'John Gileth',
'Care/dwe',
'Liuwassa',
'Ustawi',
'Nssf',
'Kilol',
'Nado',
'Judge Mchome',
'Minis',
'Milenia',
'Water User Group',
'Opec',
'Government /sda',
'Farm-africa',
'Bffs',
'Kyela-morogoro',
'Ggm',
'Msikitini',
'Kwik',
'Shelisheli Commission',
'Mungaya',
'Baptist Church',
'Tgts',
'Unknown',
'Ndorobo Tours',
'Zaben',
'Serikali Ya Kijiji',
'Enyueti',
'Watu Wa Marekani',
'Regina Group',
'Snv-swash',
'Seram',
'Lcdg',
'Adap',
'Laizer',
'African Barrick Gold',
'Salehe',
'Jumanne',
'Masai Land',
'Jipa',
'S. Kumar',
'Hpa',
'Mp Mzeru',
'W.D &',
'Wafidhi Wa Ziwa T',
'Matimbwa Sec',
"Lee Kang Pyung's Family",
'Rwsssp',
'Rural Drinking Water Supply',
'Mhoranzi',
'Woyege',
'Quick Win Project',
'Muslimu Society(shia)',
'Morovian Church',
'Grazie Franco Lucchini',
'Pankrasi',
'Irevea Sister Water',
'Unesco',
'Iucn',
'Kdc',
...]
```

In [18]:

```
df[df['funder'] == '0']
```

Out[18]:

id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
----	------------	---------------	--------	------------	-----------	-----------	----------	----------	-------------	-----	-----

id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	quality_group
145	28668	2013-03-16	0	-15	0	39.527114	6.988748	Msikitini	0	...	
152	60983	2013-03-16	0	-15	0	39.527114	6.988748	Msikitini	0	...	
393	39749	2013-03-18	0	28	0	39.159887	6.902548	Kwa Chambuso	0	...	
417	15832	2013-03-22	0	30	0	39.178404	6.938013	Ccm Kivule	0	...	
428	50233	2013-03-12	0	30	0	39.178849	6.973206	Ofisi Ya Kata	0	...	
...
59237	2138	2013-03-19	0	81	0	39.119109	6.898919	Kata	0	...	
59243	3396	2013-03-16	0	-20	0	39.524021	6.984802	Kwa Mariwala	0	...	
59276	62818	2013-03-21	0	18	0	39.183790	6.897566	Kwa Mkunduge	0	...	
59351	55322	2013-03-18	0	-19	0	39.534599	7.088183	Kwa China	0	...	
59387	26640	2013-03-12	0	25	0	39.176480	6.957098	Kwa Maliba	0	...	

777 rows x 42 columns

We saw in the value_counts that there was already a 'Not Known' value so the nulls were changed to 'Not Known'.

In [19]:

```
#replace nulls in 'funder' col with "Not Known"
df['funder'].fillna(value="Not Known", inplace=True)
```

A value of '0' was also treated as a null. These values seemed out-of-place in the column and didn't seem to describe a particular organization or individual responsible for funding.

In [20]:

```
df['funder'] = df['funder'].replace(to_replace={'0':'Not Known'}, value=None)
```

In [21]:

```
#confirm replacement of nulls
df['funder'].isna().sum()
```

Out[21]:
0

In [22]:

```
df[df['funder'] == '0']
```

Out[22]:

id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	quality_group
----	------------	---------------	--------	------------	-----------	-----------	----------	----------	-------------	-----	---------------

0 rows x 42 columns

We've confirmed removal of null values and '0' values from the 'funder' column.

installer

In [23]:

```
#organization that installed the well
df['installer'].value_counts()
```

Out[23]:

```
DWE          17402
Government   1825
RWE          1206
Commu        1060
DANIDA       1050
...
Ta           1
Zao          1
Marumbo Community  1
BATIST CHURCH  1
Regina group  1
Name: installer, Length: 2145, dtype: int64
```

Installer had 2,154 unique values which could be an issue if we planned on including this as a categorical feature in our modeling.

In [24]:

```
df['installer'].isna().sum()
```

Out[24]:

```
3655
```

There were 3,655 nulls in 'installer'. This value matched the number of nulls we saw in 'funder', which could suggest that the same rows had missing values.

In [25]:

```
list(df['installer'].unique())
```

Out[25]:

```
['Roman',
 'GRUMETI',
 'World vision',
 'UNICEF',
 'Artisan',
 'DWE',
 'DWSP',
 'Water Aid',
 'Private',
 'DANIDA',
 'Lawatefuka water sup',
 'WEDECO',
 'Danid',
 'TWE',
 'ISF',
 'Kilolo Star',
 'District council',
 'Water',
 'WU',
 nan,
 'Not known',
 'Central government',
 'CEFA',
 'Commu',
 'Accra',
 'World Vision',
 'LGA',
 'MUWSA',
 'KKKT _ Konde and DWE',
```

'Government',
'Olgilai village community',
'KKKT',
'RWE',
'Adra /Community',
'SEMA',
'SHIPO',
'HESAWA',
'ACRA',
'Community',
'IFAD',
'Sengerema Water Department',
'HE',
'ISF and TACARE',
'Kokeni',
'DA',
'Adra',
'ALLYS',
'AICT',
'KIUMA',
'CES',
'District Council',
'Ruthe',
'Adra/Community',
'Tulawaka Gold Mine',
'KKT C',
'Hesawa',
'Water board',
'LOCAL CONTRACT',
'WFP',
'LIPS',
'TASAF',
'World',
'O',
'SW',
'Shipo',
'Fini water',
'Kanisa',
'OXFARM',
'VILLAGE COUNCIL Orpha',
'Villagers',
'Idara ya maji',
'FPCT',
'WVT',
'Ir',
'DANID',
'Angli',
'secondary school',
'Amref',
'JBG',
'DADIS',
'International Aid Services',
'RW',
'Dmdd',
'TCRS',
'RC Church',
'WATER AID',
'JICA',
'Gwasco L',
'AF',
'AMREF',
'wananchi',
'FW',
'Central Government',
'MWE &',
'Gove',
'RC CHURCH',
'TDFT',
'RWE/DWE',
'Central govt',
'World Bank',
'TWESA',

'Norad',
'Hans',
'FinW',
'FIN WATER',
'OXFAM',
'Plan Internationa',
'District Council',
'RWEDWE',
'Fini Water',
'ANGLI',
'CDT',
'North',
'Oikos E .Africa',
'SHAWASA',
'UN',
'NORAD',
'Save the rain',
'John gemuta co',
'TLC',
'RC Churc',
'Plan Int',
'Phase',
'LVIA',
'Rhobi',
'Makonde water population',
'RWE/ Community',
'Is',
'KILI WATER',
'RDDC',
'FINN WATER',
'FINI WATER',
'DHV',
'Kamama',
'DDCA',
'Victoria company',
'RWSSP',
'Ce',
'KYASHA ENTERPR',
'ERETO',
'REDESO',
'Villa',
'Priva',
'KUWAIT',
'Mw',
'Magadini-Makiwaru wa',
'Dr. Matomola',
'Af',
'RCchurch/CEFA',
'Tardo',
'GOVERNMENT',
'Individuals',
'Chamavita',
'GEN',
'Missi',
'Safari Roya',
'DAWASCO',
'Gover',
'Mission',
'DWE/',
'Halmashauri ya wilaya sikonge',
'Ki',
'Rhoda',
'HAPA SINGIDA',
'Consulting Engineer',
'Karugendo',
'Co',
'Marafip',
'COSMOS ENG LTD',
'World banks',
'Tanz',
'Handeni Trunk Main(',
'SIMBA CO',

'Local technician',
'Village',
'Centr',
'CONS',
'DW',
'DCT',
'IRC',
'District water department',
'Sabodo',
'MLADE',
'I.E.C',
'LWI',
'Kiliflora',
'ICS',
'T. N. karugendo',
'DED',
'Kuwait',
'ADP',
'JWIN CO',
'BILILA',
'TPP',
'GOVER',
'CIPRO/Government',
'MWE',
'MTUWASA',
'Unisef',
'REGIONAL WATER ENGINEER ARUSHA',
'IDARA',
'Wizara ya maji',
'Tasaf and Lga',
'JAICA',
'KKKT-Dioces ya Pare',
'Onesm',
'Te',
'MTN',
'HESAWS',
'Islamic',
'Local',
'KTA C',
'RC',
'Killflora /Community',
'Distri',
'Maji block',
'CALTAZ KAHAMA',
'GOVERNME',
'Omar Ally',
'HAM',
'QUWKWIN',
'ADRA',
'DO',
'DH',
'RC Ch',
'SAXON BUILDING CONTRACTOR',
'Bokera W',
'Bulyahunlu Gold Mine',
'MBIUWASA',
'ADRA /Government',
'The Isla',
'Rotary club',
'YELL LTD',
'Care international',
'KIMKUM',
'Tanesco',
'CJEJOW CONSTRUCTION',
'Victoria',
'TLTC',
'Wachina',
'WE',
'HSW',
'Communit',
'Kibaha Town Council',
'Dr. Matobola',

'Go',
'DWR',
'Huches',
'WATERAID',
'Maswi company',
'Kiliwater',
'TA',
'wanan',
'MEM',
'Region water Department',
'Jeica',
'Ndanda missions',
'District Water Department',
'MSF/TACARE',
'Fathe',
'DARDO',
'Wa',
'MSIKIT',
'Regional Water',
'D',
'VILLAGE COUNCIL',
'RDC',
'TLC/John Majala',
'Kilwa company',
'Local technician',
'TASSAF',
'VWC',
'PIDP',
'TAN PLANT LTD',
'Japan Government',
'Kata',
'GTZ',
'ISF/Government',
'KUWASA',
'Hydrotec',
'Pr',
'Ch',
'Jaica',
'Taboma/Community',
'P',
'Ubung',
'Chur',
'BESADA',
'Action Contre La Faim',
'Wanjoda',
'CBHCC',
'HW/RC',
'Sumbaw',
'CCEC',
'Nice',
'CCT',
'World Vission',
'Inter',
'DMMD',
'WORLD BANK',
'AQUA BLUES ANGELS',
'MACK DONALD CONTRACTOR',
'Water Aid /sema',
'Henure Dema',
'Kirdep',
'ADRA/Government',
'Kilwater',
'Da',
'Villi',
'KOYI',
'AD',
'Arab community',
'District water depar',
'HOLLAND',
'RC church/Central Gover',
'Active MKM',
'GEOTAN',

'LENCH',
'NCAA',
'CHINA HENAN CONSTUCTION',
'Kaembe',
'Ma',
'FinWater',
'Kuamu',
'Adra/ Community',
'Locall technician',
'UKILIG',
'Mbunge',
'The desk and chair foundat',
'DUWAS',
'Diwani',
'Biore',
'Water aid /sema',
'KKKT CHURCH',
'EA',
'Halmashauri ya manispa tabora',
'ML appro',
'SHY BUILDERS',
'Finwater',
'JIKA',
'Orien',
'DMDD',
'DWE}',
'CDTF',
'KAEMP',
'TUWASA',
'MARAFIP',
'MDRDP',
'Jeshi la wokovu',
'kuwait',
'MBOMA',
'Grobal resource alliance',
'Village Council',
'Shamte Said',
'AUWASA',
'WSDP',
'COUN',
'KIDP',
'Mombo urban water s',
'TRIDEP',
'Wananchi',
'Martha Emanuel',
'St',
'GIDA contractor',
'WD and ID',
'Padep',
'Po',
'Village Counil',
'MINISTRY OF WATER',
'Ga',
'K',
'Swiss If',
'Miziriol',
'Yasini Selemani',
'DBSPE',
'European Union',
'H',
'TPP TRUSTMOSHI',
'Atisan',
'Jika',
'ISF/TACARE',
'Oikos E.Africa',
'Hydom Luthelani',
'Kalumbwa',
'ILCT',
'MS',
'RUVUMA BASIN',
'Gold star',
'Mi',

'Mzungu Paul',
'Kanisa katoliki',
'Caltas',
'RED CROSS',
'World bank',
'Losaa-Kia water supp',
'Jica',
'PET',
'Finland Government',
'GAICA',
'Institution',
'TCRS/TLC',
'Loliondo Parish',
'GACHUMA CONSTRUCTION',
'Diocese of Geita',
'Villages',
'MSABI',
'Total landcare',
'VICTORIA DRILL CO',
'U.S.A',
'VTECOS',
'COW',
'Vill',
'Contr',
'Wadeco',
'KIM KIM CONSTRUCTION',
'Msabi',
'VC',
'CMSR',
'Ko',
'Roman Catholic Rulenge Diocese',
'Shule',
'W',
'inkinda',
'Africa Amini Alama',
'Consultant',
'L',
'Moroil',
'Sekei village community',
'US Embassy',
'PIT COOPERATION LTD',
'Do',
'world',
'Government /TCRS',
'UNHCR',
'DESK C',
'Dr.Matomola',
'FOLAC',
'Village govt',
'BSF',
'Roman Cathoric Same',
'RWE/Community',
'Mileniam project',
'ACTIVE TANK CO',
'Ncaa',
'Africa Islamic Agency Tanzania',
'Max Mbise',
'DADS',
'Institutional',
'SOWASA',
'CCPK',
'AUSTRALIA',
'not known',
'Kalago enterprises Co.Ltd',
'Roman Catholic',
'NANRA contractor',
'WORLD VISION',
'No',
'ADP Busangi',
'TSRC',
'SOLIDAME',
'Barry A. Murphy',

'Tanzania Government',
'WILLIAMSON DIAMOND LTD',
'TAG',
'The I',
'Total Landcare',
'CENTRAL GOVERNMENT',
'Arabs Community',
'Secondary school',
'Water Aid/Sema',
'Jiks',
'Konoike',
'ABASIA',
'LAMP',
'SINGIDA YETU',
'RWSP',
'MDALA Contractor',
'Netherlands',
'DWT',
'TCRS /CARE',
'Makonde',
'Japan',
'Milenium',
'Goldstar',
'District COUNCIL',
'MUWASA',
'Green',
'Kigoma municipal',
'KINAPA',
'CHINA HENAN CONTRACTOR',
'Musa',
'TANAPA',
'Ministry of water engineer',
'EFG',
'MASWI',
'Kyariga',
'Roman Cathoric -Kilomeni',
'Mbozi Secondary School',
'TASAF/DMDD',
'MWS',
'Roman catholic',
'Shekhe',
'Rished',
'KONOIKE',
'Pata',
'TAHEA',
'Luthe',
'Kalta',
'Pentecost church',
'Amboni Plantation',
'Municipal',
'Sekondari',
'Kalitasi',
'HOTELS AND LOGGS TZ LTD',
'DISTRICT COUNCIL',
'Germany',
'Orphanage',
'WWF',
'W.B',
'IDYDC',
'SIA Ltd',
'WINAM CONSTRUCTION',
'RIDEP',
'NORA',
'SCHOOL',
'Village community',
'British',
'Msuba',
'Villaers',
'TLC/Thimotheo Masunga',
'WB',
'Council',
'DAK',

'COCANE',
'WINAMU CO',
'Ubalози wa Marekani',
'Conce',
'BGM',
'DMK',
'Mviwa',
'KA',
'MGM',
'AIMGOLD',
'YEBE CHIKOMESH',
'Omari Mzee',
'Petro Patrice',
'Camartec',
'Total land care',
'Wasso companies',
'DASP',
'Islamic Agency Tanzania',
'Tanz Egypt technical coopera',
'Village Govt',
'local technician',
'TAWASA',
'WATER AID',
'AAR',
'MSF',
'Di',
'Mackd',
'MAMAD',
'PADEP',
'Fabia',
'CONCERN',
'ITALI',
'Water aid/sema',
'Save the rain USA',
'Plan Tanzania',
'Roman Church',
'Singasinga',
'RC/Mission',
'In',
'V',
'Korogwe water works',
'PCI',
'Atlas',
'DWE /TASSAF',
'Local te',
'World Division',
'Gwaseco',
'Kambi Migoko',
'AI',
'SAUWASA',
'Nyakilanganyi',
'DEE',
'MANYARA CONSTRUCTION',
'Rotte',
'KMCL',
'LINDALA CO',
'Government /Community',
'CCPS',
'SI',
'Rundu man',
'Water Aid/sema',
'Naishu construction co. ltd',
'WOULD BANK',
'Mark',
'Cosmo',
'Halmashauri',
'Concern /government',
'Quick win project',
'Mh Kapuya',
'Halmashauri ya wilaya',
'Edward',
'COMMU',

'Baric',
'Consuting Engineer',
'JANDU PLUMBER CO',
'FiNI WATER',
'CPRO',
'Getekwe',
'Jicks',
'Wahidi',
'Mohamed Ally',
'ASDP',
'CITIZEN ENGINE',
'KADP',
'Dar es salaam Technician',
'Halmashauli',
'ACORD',
'MA',
'Water Aid/Sema',
'RC church/CEFA',
'Wedeco',
'DWE/Ubalozzi wa Marekani',
'VIFAFAI',
'Kwasenenge Group',
'Cosmos Engineering',
'OLDONYOLENGAI',
'NYAKILANGANI CO',
'Village Community',
'MINJINGU',
'EL',
'Songa',
'Consultant and DWE',
'AC',
'Gain',
'DASIP',
'TANROAD',
'Tasaf',
'Wasso',
'Teonas Wambura',
'Mgaya Masese',
'TUKWALE ENTERP',
'Sao',
'MWAKI CONTRACTOR',
'VIEN CONSTRUCTION',
'mwita kichere',
'DADS/village community',
'Africare',
'Mosque',
'Chiko',
'central government',
'VITECOS',
'IN',
'Msikiti',
'Word Bank',
'Kwamdulu estate',
'SEMA Consultant',
'Concern',
'Belgiam Government',
'Wanan',
'Exaud Msambwa',
'Niger',
'MWANZA',
'SONGAS',
'MINISTRYOF WATER',
'COMMUNITY',
'Zaburi and neighbors',
'NDM',
'Killflora/ Community',
'PART',
'secondary',
"lion's club",
'lutheran church',
'Mileniam',
'UYOGE',

'Christina Magoge',
'Canada na Tanzania',
'FRANKFURT',
'GOVERN',
'Kuji foundation',
'Mamvua Kakungu',
'Rusumo Game reserve',
'MTUWASA and Community',
'ACT MARA',
'UMOJA DRILLING',
'KkKT',
'SDA',
'Mzinga A',
'RE',
'LOOCIP',
'SUA',
'RUNDAGA',
'RWE /Community',
'Wo',
'Happy watoto foundation',
'GDP',
'VILLAGE COUNCIL',
'MBULU DISTRICT COUNCIL',
'Maliasili',
'Roman Ca',
'NZILA',
'stansilaus',
'AFRICAN DEVELOPMENT FOUNDATION',
'FPTC',
'KARUMBA BIULDING COMPANY LTD',
'Kalugendo',
'Village Government',
'Tabraki',
'MASWI DRILLING',
'Ikela Wa',
'Shallow well',
'WEDECO/WESSONS',
'CIPRO/CARE/TCRS',
'Wasso contractors',
'villagers',
'Mwanga town water authority',
'Jumanne Siabo',
'Mama Kalage',
'Hindu',
'Rural',
'TANAP',
'Makonde water supply',
'villigers',
'Bingo foundation Germany',
'Ilwilo community',
'St ph',
'WDECO',
'LIVI',
'Pet Corporation Ltd',
'DWE & LWI',
'LC',
'KKKT Leguruki',
'HIAP',
'Matyenye',
'DIMON',
'Italy government',
'MASWI DRILL',
'WVC',
'TACRI',
'Hasnein Muri',
'SIMBA LODGE',
'Faudh Tamimu',
'Free Pentecoste Church of Tanz',
'Summit for water/Community',
'Sanje Wa',
'Makundya',
'Individual',

'OLA',
'RC C',
'TREDEP',
'Consultant Engineer',
'AQUA WEL',
'Cental Government',
'Nyanza road',
'Kizenga',
'KKT',
'HAPA',
'Oikos E. Africa',
'Ramadhani Nyambizi',
'Mdala Contractor',
'DENISH',
'Mkuyu',
'GOVERN',
'GACHUMA GINERY',
'Resolute',
'Morrov',
'Serikali ya kijiji',
'Counc',
'Igolola community',
'S',
'NYAKILANGANI CONSTRUCTION',
'RDWS',
'Said Omari',
'AFRICA MUSLIM',
'IADO',
'W/',
'Ngiresi village community',
'UDC/Sema',
'AMP contractor',
'rc ch',
'QWICKWIN',
'Mwita Muremi',
'TLC/Samora',
'Oikos E.Afrika',
'Ruangwa contractor',
'HAYDOM LUTHERAN HOSPITAL',
'VICFISH LTD',
'Lindi contractor',
'RC CH',
'Kilomber',
'Pet Coporation Ltd',
'Afroz Ismail',
'Ja',
'commu',
'Sisal Estste Hale',
'KOREA',
'CVS Miss',
'Songas',
'Living water international',
'Kajima',
'Missio',
'UAACC',
'GERMANY MISSIONARY',
'MI',
'Rips',
'LVA Ltd',
'BUKUMB',
'Taasi',
'STAMPERS',
'Meru Concrete',
'WIZARA',
'MLAKI CO',
'Segera Estate',
'WADECO',
'Hospi',
'Cebtral Government',
'local technician',
'Siza Mayengo',
'SAXON',

'Greec',
'KASHERE',
'GURUMETI SAGITA CO',
'China',
'MP',
'Islam',
'water board',
'AMP Contract',
'Thomasi busigaye',
'Local technitian',
'SIJM',
'KKKT Ndrumangeni',
'YUMBAKA ENGINEERING',
'TAG CHURCH',
'Usambala sisters',
'KOBORG Contractor',
'hesawa',
'Water Authority',
'Mr Chi',
'Hearts helping hands.Inc.',
'IDEA',
'Selous G',
'SULEMAN IDD',
'Pump entecostal Sweeden',
'Ester Ndege',
'Nyabarongo Kegoro',
'Canop',
'QUIK',
'DADP',
'Kanisani',
'CARTAS',
'Mzung',
'wizara ya maji',
'VILLAGE COUNCIL .ODA',
'CG',
'Caltus',
'Cons',
'ISSAC MOLLEL',
'malola',
'DCCA',
'Juma Maro',
'Water Project Mbawala chini',
'Unicef',
'Totoland care',
'Maswi drilling co ltd',
'NDDP',
'KMT',
'NGINIL',
'Serengeti District concil',
'RC church',
'VILLAG',
'Local technical tec',
'Cultus',
'T',
'Hery',
'OBC',
'RUDEP',
'RWE Community',
'Nyamongo Gold mining',
'Redep',
'Norani',
'Mahita',
'-',
'Villag',
'germany',
'KARUMBA BIULDIN',
'AIXOS',
'Selikali',
'DDP',
'Village government',
'Zacharia MTN',
'Africa',

'PAD',
'KASHWA',
'TWEENDE PAMOJA',
'Uhai wa mama na mtoto',
'OLOMOLOKI',
'Ardhi water well',
'Distric Water Department',
'gwitembe',
'Conta',
'HOWARD HUMFREYS',
'SHUWASA',
'JANDU PLUMBER CO',
'Makori',
'Sangea District Coun',
'CHINA',
'British colonial government',
'Maendeleo ya jamii',
'CARITAS',
'Taes',
'KWIKWIZ',
'SEMA CO LTD',
'SENAPA',
'REGWA COMPANY OF EGYPT',
'COBASHEC',
'AQUA Wat',
'Dr.Matobola',
'Central basin',
'Mamlaka ya maji ngara',
'PRF',
'Church',
'Magadini Makiwaru wa',
'Mpang',
'KAYEMPU LTD',
'TRACHOMA',
'FURAHIA TRADING',
'HESAW',
'Moravian',
'Samsoni',
'MD',
'GURUMETI SAGITA',
'Songea District Coun',
'Cast',
'N.P.R.',
'Panone',
'Hemed Abdallah',
'Lawate fuka water su',
'St Gasper',
'WINNIN SPIRIT CO',
'Ha',
'MMG GOLD MINE',
'P.N.R.',
'Nandra Construction',
'Mchuk',
'African Realief Committe of Ku',
'SCOTT',
'D\$L',
'Mzungu',
'Vi',
'JLH CO LTD',
'Msiki',
'Namungo',
'Nassor Fehed',
'TWESA /Community',
'DBFPE',
'EF',
'Serikali',
'Mgaya Mwita',
'Clause workers',
'MLAKI CO',
'Busoga trust',
'mzee mabena',
'NORAD/',

```
'BR',
'local technician',
'Community',
'Brad',
'Tanganyika Basin',
'MORNING CONSTRUCTION',
'Healt',
'Governme',
'Roma',
'KUMKUM',
'PNR co',
'Muslims',
'Paffec',
'Tansi',
'CRAELIUS',
'APM',
'Zao water spring X',
'TASA',
'CSPD',
'CIPRO/CARE',
'DALDO',
'VIFAF',
'MTC',
'TCRS Kibondo',
'Howard and humfrey consultant',
'RUDEP/',
'LUNGWE',
'Dhinu',
'AIC KI',
'Mataro',
'FINI Water',
'Mombo urban water',
'REDAP',
'Kagulo',
'TMP',
...]
```

Again, we saw there was a 'Not known' value in 'installer', which we decided to use to replace nulls.

In [26]:

```
df['installer'].fillna(value='Not known', inplace=True)
```

In [27]:

```
df['installer'].isna().sum()
```

Out[27]:

```
0
```

We noticed significant overlap in values for 'funder' and 'installer'. From our business understanding, this made sense as organizations who were funding water pumps were also likely to be responsible for their installation.

In [28]:

```
#check cases where the installer is not also the funder
df[df['funder'] != df['installer']]
```

Out[28]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...
1	8776	0.0	2013-03-06	Grumeti	1399	GRUMETI	34.698766	-2.147466	Zahanati	0	...
2	34310	25.0	2013-02-25	Lottery Club	686	World vision	37.460664	-3.821329	Kwa Mahundi	0	...
3	67743	0.0	2013-01-28	Unicef	263	UNICEF	38.486161	-	Zahanati	0	...

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...
4	19728	0.0	2011-07-13	Action In A	0	Artisan	31.130847	-1.825359	Shuleni	0	...
5	9944	20.0	2011-03-13	Mkinga Distric Coun	0	DWE	39.172796	-4.765587	Tajiri	0	...
...
59395	60739	10.0	2013-05-03	Germany Republi	1210	CES	37.169807	-3.253847	Area Three Namba 27	0	...
59396	27263	4700.0	2011-05-07	Cefa-njombe	1212	Cefa	35.249991	-9.070629	Kwa Yahona Kuvala	0	...
59397	37057	0.0	2011-04-11	Not Known	0	Not known	34.017087	-8.750434	Mashine	0	...
59398	31282	0.0	2011-03-08	Malec	0	Musa	35.861315	-6.378573	Mshoro	0	...
59399	26348	0.0	2011-03-23	World Bank	191	World	38.104048	-6.747464	Kwa Mzee Lugawa	0	...

54481 rows x 42 columns



In [29]:

```
#investiating funder and installer relationship with World Bank as a specific value case
df[df['installer']=='World Bank']
```

Out[29]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
280	9474	250.0	2013-02-12	World Bank	1343	World Bank	30.017098	-4.332583	Jeshini	0	...	
304	43345	0.0	2012-10-23	World Bank	0	World Bank	33.430917	-4.389084	Shule Ya Msingi	0	...	
2071	28588	0.0	2012-10-23	World Bank	0	World Bank	33.436073	-4.421944	Mwanza Road	0	...	
5699	74077	0.0	2012-10-18	World Bank	0	World Bank	33.435268	-4.671744	Kwa Fupe	0	...	
6327	44441	0.0	2012-10-12	World Bank	0	World Bank	33.150261	-3.705625	Kwa Maraba	0	...	
...	
58348	54574	0.0	2012-10-23	World Bank	0	World Bank	33.442457	-4.426592	Kilabuni	0	...	
58434	54545	250.0	2013-02-12	World Bank	1311	World Bank	30.017351	-4.323850	Mission	0	...	
58681	6335	250.0	2013-02-12	World Bank	1306	World Bank	30.017716	-4.308701	Kwa Thomas	0	...	
58691	67819	0.0	2012-10-12	World Bank	0	World Bank	33.142605	-3.699442	Kwa Nyamizi Maswa	0	...	
58803	44356	0.0	2013-02-06	World Bank	1697	World Bank	29.784471	-4.445344	Kwa Esrom	0	...	

95 rows x 42 columns



In [30]:

```
df[(df['installer']=='World') & (df['funder']!='World Bank')]
```

Out[30]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
33726	63910	0.0	2011-04-02	Nethe	141	World	38.204463	-6.870355	Kwa Kiwele	0	...	

1 rows x 42 columns

In [31]:

```
df[df['installer'] == 'Not known']
```

Out[31]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
34	41583	0.0	2011-02-23	Not Known	-41	Not known	39.812912	-7.889986	Msikitini Wa Ijumaa	0	...	
35	57355	0.0	2013-03-28	Not Known	1546	Not known	36.618699	-3.293003	Sekondari	0	...	
43	19282	0.0	2013-01-15	Not Known	1642	Not known	34.967789	-4.628921	Mvae Primary	0	...	
47	13620	0.0	2011-07-27	Not Known	0	Not known	33.540607	-9.172905	Mahakamani	0	...	
65	51072	0.0	2013-02-09	Not Known	1415	Not known	34.621598	-5.173136	Nyambi	0	...	
...
59357	46563	0.0	2013-02-19	Not Known	1635	Not known	34.971841	-5.098362	Shabani	0	...	
59366	55232	0.0	2013-02-02	Not Known	1541	Not known	34.765729	-5.027725	Joshoni	0	...	
59370	14796	200.0	2013-01-29	Not Known	1154	Not known	30.058731	-4.902633	Village Office	0	...	
59376	34716	0.0	2013-02-03	Not Known	1581	Not known	34.821039	-5.076258	Nasingo	0	...	
59397	37057	0.0	2011-04-11	Not Known	0	Not known	34.017087	-8.750434	Mashine	0	...	

3672 rows x 42 columns

In [32]:

```
df['installer'].value_counts(normalize=True)[:20]
```

Out[32]:

DWE	0.292963
Not known	0.061818
Government	0.030724
RWE	0.020303
Commu	0.017845
DANIDA	0.017677
KKKT	0.015118
Hesawa	0.014141
0	0.013081
TCRS	0.011902
Central government	0.010471
CES	0.010269

```
Community      0.009310
DANID          0.009293
District Council 0.009276
HESAWA         0.009074
LGA            0.006869
World vision   0.006869
WEDECO         0.006684
TASAF          0.006667
Name: installer, dtype: float64
```

As with the 'funder' column, we treated values of '0' as nulls and replaced them with 'Not known'.

In [33]:

```
df['installer'] = df['installer'].replace(to_replace={'0': 'Not known'}, value=None)
```

In [34]:

```
df['installer'].value_counts(normalize=True)[:20]
```

Out[34]:

```
DWE      0.292963
Not known 0.074899
Government 0.030724
RWE      0.020303
Commu    0.017845
DANIDA   0.017677
KKKT     0.015118
Hesawa   0.014141
TCRS     0.011902
Central government 0.010471
CES      0.010269
Community 0.009310
DANID    0.009293
District Council 0.009276
HESAWA   0.009074
World vision 0.006869
LGA      0.006869
WEDECO   0.006684
TASAF    0.006667
District council 0.006599
Name: installer, dtype: float64
```

In [35]:

```
df[df['installer'] == "0"]
```

Out[35]:

id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	quality_group
0 rows x 42 columns											

subvillage

In [36]:

```
#geographic location
df['subvillage'].isna().sum()
```

Out[36]:

371

In [37]:

```
df['subvillage'].value_counts()
```

```
Out[37]:
```

```
Madukani      508
Shuleni       506
Majengo       502
Kati          373
Mtakuja       262
...
Masaladi      1
Mgodi         1
Ipwasi        1
Tobo         1
Bombambili 2  1
Name: subvillage, Length: 19287, dtype: int64
```

There were 3,635 nulls in 'subvillage'. This column has over 19,287 unique values.

```
In [38]:
```

```
list(df['subvillage'].unique())
```

```
Out[38]:
```

```
['Mnyusi B',
 'Nyamara',
 'Majengo',
 'Mahakamani',
 'Kyanyamisa',
 'Moa/Mwereme',
 'Ishinabulandi',
 'Nyawishi Center',
 'Imalauduki',
 'Mkonomre',
 'Mizugo',
 'Ngondombwito',
 'Nkilifa',
 'Omarini',
 'Mwabasabi',
 'Tunzi',
 'Kidudumo',
 'Yeriko',
 'Center',
 'Manyanya',
 'Ibabachegu',
 'Mkanivega',
 'Mkonga Juu',
 'Msasa',
 'Kitereni',
 'Shuleni',
 'Chakahaya',
 'Kiyao',
 'Merali',
 'Karume',
 'Kudipera',
 'Mosheni',
 'Lupanga A',
 'Kilombero B',
 'Afya',
 'Ndanganyika',
 'Baura',
 'Mwanzala',
 'Nyabwai B',
 'Reli B',
 'Kilunduwe',
 'Sokoni',
 'Mwarufyu',
 'Marurani Juu',
 'Isenegeja',
 'Kachulu',
 'Mpandapanda',
 'Mlandege Juu',
 ...]
```


'Ikanga',
'Msaranga Street',
'Maporomoko',
'Isimba',
'Kagoye B',
'Dhobi Street',
'Msufini',
'Soyekiutu',
'Ants B',
'Izimbya 'A'',
'Nairobi',
'Nkaloi',
'Kiganza Centre',
'Ulkusare',
'Mzimba',
'Mfumbu',
'Mahalule',
'Kihanga',
'Iponda',
'Kisoro',
'Mtakuja',
'Lembuka',
'Mapinduzi',
'Kalimungoma',
'Umkituri',
'Usita',
'Komoro',
'Kulasi Majengo',
'Butimba',
'Ikovo',
'Dongo',
'Moivaro',
'Bushoma',
'Shule',
'Bulyahilu Center B',
'Mlanda B',
'Kasharunga',
'Magwila',
'Juhudi',
'Kilundo',
'Mwena',
'Mara B',
'Mushasha',
'Kitobo',
'Kishiha',
'Misasi C',
'Msewo Mwaazi',
'Madukani A',
'Kiruku Mchangamweupe',
'Mashine',
'Mlima Ndabaneze',
'Nguvumali',
'Tema',
'Kwa Nyange',
'Malula',
'Maendeleo',
'Kidete',
'Ipuguso',
'Mseseweni',
'Kiwawa',
'Nyansalala',
'Vimetu',
'Chang'Ombe",
'Kalamila',
'Mingo',
'Kabarongo',
'Mwebebonda',
'Nendebe',
'Ihela Shu',
'Ikongora',
'Majengo B',
'Ewerendeke',

'Busekele',
'Kitega Uchumi',
'Mosi',
'Utengule',
'Butondolo',
'Isera',
'Nyarutembo',
'Msumbiji',
'Wihanga',
'Igalula A',
'Mgwashi',
'Ligelango',
'Ipumpila',
'Mkopwe',
'Nyakafundikwa A',
'Wichamoyo',
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'Kilimbili',
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'Mgaraganza',
'Sokomoko',
'Mondelo',
'Mifugo',
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'Wangama',
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'Itubula',
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'Ibambula',
'Balanga',
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'Bwiti',
'Station',
'Guanaya',
'Mhandu',
'Uhuru',
'Mayadi',
'Imalambeo',
'Tanya',
'Kitunda',
'Mwasubi',
'Tundu La Leo',
'Mkuya',
'Bwelu A',
'Kitundu',
'Yaniko B',
'Sandege',
'Kituntu A',
'Idwele',
'

```
'Omukigando',
'Mbalizi',
'Ilelamhina Shuleni',
'Kaija',
'Busabaga',
'Kizinga A',
'Ghalani',
'Lyabasura',
'Uparo',
'Kabasonge',
'Wiligwamabu',
'Sinai',
'Mihogoni',
'Gongoti',
'Mwegerezi',
'Mawanda',
...]
```

In [39]:

```
df[df['subvillage'] == 'Unknown']
```

Out[39]:

id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	quality_group
0 rows x 42 columns											

Unlike our previous cases, there wasn't a clear value to change nulls to. We decided to create a new 'Unknown' value to replace nulls rather than dropping these rows.

In [40]:

```
df['subvillage'].fillna(value='Unknown', inplace=True)
```

In [41]:

```
df['subvillage'].isna().sum()
```

Out[41]:

0

public_meeting

In [42]:

```
df['public_meeting'].isna().sum()
```

Out[42]:

3334

In [43]:

```
df['public_meeting'].value_counts()
```

Out[43]:

```
True      51011
False     5055
Name: public_meeting, dtype: int64
```

There were 3,334 nulls in 'public_meeting'. Given the large imbalance between True and False value counts, we decided that nulls should match the majority class, True.

In [44]:

```
df['public_meeting'].fillna(value=True, inplace=True)
```

In [45]:

```
df['public_meeting'].isna().sum()
```

Out[45]:

0

scheme_management

In [46]:

```
#who operates the waterpoint (organization/category)
df['scheme_management'].isna().sum()
```

Out[46]:

3877

There were 3,877 nulls in 'scheme_management'.

In [47]:

```
df['scheme_management'].value_counts()
```

Out[47]:

```
VWC          36793
WUG           5206
Water authority  3153
WUA           2883
Water Board   2748
Parastatal    1680
Private operator 1063
Company       1061
Other         766
SWC            97
Trust         72
None           1
Name: scheme_management, dtype: int64
```

In [48]:

```
#investigating the one value_count of None
df[df['scheme_management'] == "None"]
```

Out[48]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qu
23603	23849	50.0	2013-03-18	Not Known	-11	Not known	39.431194	-7.100783	Kwa Nyamtawa	0	...	

1 rows x 42 columns



In [49]:

```
df[df['funder'] == '0']
```

Out[49]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	quality_group
--	----	------------	---------------	--------	------------	-----------	-----------	----------	----------	-------------	-----	---------------

0 rows x 42 columns



Since there was already an instance of 'None' in this column, we changed all nulls to 'None'.

In [50]:

```
df['scheme_management'].fillna(value='None', inplace=True)
```

In [51]:

```
df['scheme_management'].isna().sum()
```

Out[51]:

0

scheme_name

This column had a large number of nulls: 28,166. Since the values appeared to be individual names with great diversity, we decided this column was unusable for our modeling. However, we didn't drop it from the dataframe since our modeling process uses a pipeline for this purpose.

In [52]:

```
#who operates the waterpoint  
#how is this information different from scheme_management?  
df['scheme_name'].value_counts()
```

Out[52]:

K	682
None	644
Borehole	546
Chalinze wate	405
M	400
...	
Nkwe	1
BUWADA	1
BL Bonifas Kingu	1
BL Losei	1
DMK:Anglican church	1

Name: scheme_name, Length: 2696, dtype: int64

In [53]:

```
df['scheme_name'].isna().sum()
```

Out[53]:

28166

In [54]:

```
unique_scheme = list(df['scheme_name'].unique())  
print(len(unique_scheme))  
unique_scheme
```

2697

Out[54]:

```
['Roman',  
 nan,  
 'Nyumba ya mungu pipe scheme',  
 'Zingibali',  
 'BL Bondeni',  
 'None',  
 "wanging'ombe water supply s",  
 'Makanj',  
 'Kidabu',  
 'Mashangwi',  
 'Quick wins Program',  
 'Komaka mandaka',  
 'Schodo Borehole Scheme']
```


'Borehole Borehole Scheme',
'Kitukuni water supply',
'BL Mwakikoti',
'Chalinze wate',
'Mae pipeline',
'UNDP',
'Ngana water supplied scheme',
'Itun',
'Bomala',
'Kirua kahe pumping water trust',
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'Tingi water supply',
'Mgun',
'IINIHA nined scheme'

```

'Ilolangulu water supply',
'Mwembe Water Supply',
'Kidu',
'Isapul',
'Mbimbi water gravity scheme',
'Mtangashari',
'Kishuro water sup',
'Arisi/himo',
'Zigi',
'Ndimira water supply',
'Saibala gravity water supply',
'Lihimalyao water supply',
'Mzinga r',
'Utweve',
'Kagenyi water sup',
'BL Kitahemo',
'Bwambo Water Supply',
'Njomlole water gravity scheme',
'Sauwasa water supply',
'Tagame',
'Matala pipeline',
'Gedamar water supply',
'Marera water supply',
'Kilesi water supply',
'Maka',
'Kolo',
'Kiroriko water supply',
'Malambo water scheme',
'Muungano',
'BL Nshere Juu',
'Mkata ri',
'Vunta water supply',
'Luwamakaa branch line',
'Msjimingi',
'Kitowo',
'BL Zahanati',
'Ichonde',
'Kashishi water supply',
'EKTM 2 water suply',
'Nyamasenene Water Supply',
'Rumashi gravity water supply',
'BL Sanya Hoi',
'Mangola pipe scheme',
'Chol',
'Kaguruka Water Supply',
'Mradi wa maji Kowak',
'Ibih',
'Mban',
'TM part Three water supply',
'Kabaheshi pring source',
'Makiyui stream',
...]
```

We replaced the nulls with 'None'. Although there is a strong case to drop this column (which now contains 28,810 'None' values), we left it in the dataframe.

In [55]:

```
df['scheme_name'].fillna(value='None', inplace = True)
```

In [56]:

```
df['scheme_name'].isna().sum()
```

Out[56]:

0

In [57]:

```
df['scheme_name'].value_counts()
```

```
Out[57]:
```

```
None          28810
K              682
Borehole       546
Chalinze wate  405
M              400
...
Nkwe           1
BUWADA         1
BL Bonifas Kingu  1
BL Losei       1
DMK:Anglican church  1
Name: scheme_name, Length: 2696, dtype: int64
```

Permit

```
In [58]:
```

```
df['permit'].isna().sum()
```

```
Out[58]:
```

```
3056
```

There were 3,056 nulls in 'permit'.

We assumed that if a permit status is unknown, there is no permit. The nulls were changed to False.

```
In [59]:
```

```
#if the waterpoint is permitted
df['permit'].value_counts()
```

```
Out[59]:
```

```
True      38852
False     17492
Name: permit, dtype: int64
```

```
In [60]:
```

```
df['permit'].isna().sum()
```

```
Out[60]:
```

```
3056
```

```
In [61]:
```

```
df['permit'].fillna(value=False, inplace=True)
```

```
In [62]:
```

```
df['permit'].isna().sum()
```

```
Out[62]:
```

```
0
```

Column Exploration

Now that null values have been treated, we continued our exploration of the data by examining columns individually.

```
In [63]:
```

```
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 59400 entries, 0 to 59399
Data columns (total 42 columns):
#   Column                                Non-Null Count  Dtype
---  ---                                -
0   id                                    59400 non-null  int64
1   amount_tsh                           59400 non-null  float64
2   date_recorded                        59400 non-null  object
3   funder                               59400 non-null  object
4   gps_height                           59400 non-null  int64
5   installer                            59400 non-null  object
6   longitude                            59400 non-null  float64
7   latitude                             59400 non-null  float64
8   wpt_name                             59400 non-null  object
9   num_private                          59400 non-null  int64
10  basin                                59400 non-null  object
11  subvillage                           59400 non-null  object
12  region                               59400 non-null  object
13  region_code                          59400 non-null  int64
14  district_code                       59400 non-null  int64
15  lga                                   59400 non-null  object
16  ward                                 59400 non-null  object
17  population                           59400 non-null  int64
18  public_meeting                       59400 non-null  bool
19  recorded_by                          59400 non-null  object
20  scheme_management                    59400 non-null  object
21  scheme_name                          59400 non-null  object
22  permit                               59400 non-null  bool
23  construction_year                   59400 non-null  int64
24  extraction_type                     59400 non-null  object
25  extraction_type_group                59400 non-null  object
26  extraction_type_class                59400 non-null  object
27  management                           59400 non-null  object
28  management_group                    59400 non-null  object
29  payment                             59400 non-null  object
30  payment_type                         59400 non-null  object
31  water_quality                        59400 non-null  object
32  quality_group                        59400 non-null  object
33  quantity                             59400 non-null  object
34  quantity_group                       59400 non-null  object
35  source                              59400 non-null  object
36  source_type                          59400 non-null  object
37  source_class                         59400 non-null  object
38  waterpoint_type                      59400 non-null  object
39  waterpoint_type_group                59400 non-null  object
40  id_label                             59400 non-null  int64
41  status_group                         59400 non-null  object
dtypes: bool(2), float64(3), int64(8), object(29)
memory usage: 18.2+ MB

```

In [64]:

```

#total static head (amount of water available to waterpoint)
df['amount_tsh'].value_counts()

```

Out[64]:

```

0.0          41639
500.0         3102
50.0          2472
1000.0        1488
20.0          1463
...
8500.0         1
6300.0         1
220.0          1
138000.0        1
12.0           1
Name: amount_tsh, Length: 98, dtype: int64

```

In [65]:

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

'date_recorded' is stored in yy-mm-dd format. We thought about using this column along with 'construction_year' to determine pump age. However, 'construction_year' has too many 0 values for us to construct a meaningful age feature.

```
df['construction_year'].value_counts()
```

Out[68]:

0	20709
2010	2645
2008	2613
2009	2533
2000	2091
2007	1587
2006	1471
2003	1286
2011	1256
2004	1123
2012	1084
2002	1075
1978	1037
1995	1014
2005	1011
1999	979
1998	966
1990	954
1985	945
1980	811
1996	811
1984	779
1982	744
1994	738
1972	708
1974	676
1997	644
1992	640
1993	608
2001	540
1988	521
1983	488
1975	424

```

1975      437
1986      434
1976      414
1970      411
1991      324
1989      316
1987      302
1981      238
1977      202
1979      192
1973      184
2013      176
1971      145
1960      102
1967       88
1963       85
1968       77
1969       59
1964       40
1962       30
1961       21
1965       19
1966       17
Name: construction_year, dtype: int64

```

In [69]:

```

#altitude of the well
df['gps_height']

```

Out[69]:

```

0          1390
1          1399
2           686
3           263
4             0
...
59395      1210
59396      1212
59397         0
59398         0
59399        191
Name: gps_height, Length: 59400, dtype: int64

```

In [70]:

```

df['longitude'].value_counts()

```

Out[70]:

```

0.000000      1812
37.540901         2
33.010510         2
39.093484         2
32.972719         2
...
37.579803         1
33.196490         1
34.017119         1
33.788326         1
30.163579         1
Name: longitude, Length: 57516, dtype: int64

```

In [71]:

```

#remove filler values in longitude and latitude
# dropping longitude of 0
df = df.loc[df["longitude"] != 0]
df.longitude.value_counts()

```

Out[71]:

```

22  000217      2

```



```

33.090347      2
32.982698      2
37.297680      2
33.010510      2
39.093484      2
..
37.579803      1
33.196490      1
34.017119      1
33.788326      1
35.005922      1
Name: longitude, Length: 57515, dtype: int64

```

In [72]:

```
df['latitude'].value_counts()
```

Out[72]:

```

-2.496459      2
-6.964258      2
-6.981884      2
-7.175174      2
-7.104625      2
..
-5.726001      1
-9.646831      1
-8.124530      1
-2.535985      1
-2.598965      1
Name: latitude, Length: 57516, dtype: int64

```

Both latitude and longitude seem to have placeholder values. These values are 0.000000 for 'longitude' and -2.000000e-08 for 'latitude'. There are 1,812 of these placeholders.

In [73]:

```

#name of the waterpoint
df['wpt_name']

```

Out[73]:

```

0          none
1      Zahanati
2      Kwa Mahundi
3  Zahanati Ya Nanyumbu
4      Shuleni
...
59395  Area Three Namba 27
59396  Kwa Yahona Kuvala
59397  Mashine
59398  Mshoro
59399  Kwa Mzee Lugawa
Name: wpt_name, Length: 57588, dtype: object

```

In [74]:

```
df['wpt_name'].value_counts()
```

Out[74]:

```

none          3492
Shuleni       1734
Zahanati      814
Msikitini     533
Kanisani      322
...
Kwa Kadungu   1
Nyanguruguru 1
Kwa Piguli    1
Kalakabanga   1
Kwa Mtunga Lyimo 1
Name: wpt_name, Length: 36720, dtype: int64

```

We thought 'wpt_name' is probably not a useful feature for classification, similar to 'scheme_name'. The many unique values would limit out model's performance with 37,400 unique values in this column. Some of these are subvillage names (like Shuleni), explaining why some values have much higher value counts than others.

In [75]:

```
#no context
df['num_private'].value_counts()
```

Out[75]:

```
0      56831
6        81
1        73
5        46
8        46
...
180      1
213      1
23       1
55       1
94       1
Name: num_private, Length: 65, dtype: int64
```

There was no context available for 'num_private', not even from the data source. We decided this was another column that would go unused in modeling since we were unable to identify its meaning. Additionally, the large value counts imbalance was worrying without any business understanding of these values.

In [76]:

```
#Geographic water basin
df['basin'].value_counts()
```

Out[76]:

```
Pangani      8940
Lake Victoria 8535
Rufiji       7976
Internal     7785
Lake Tanganyika 6333
Wami / Ruvu  5987
Lake Nyasa   5085
Ruvuma / Southern Coast 4493
Lake Rukwa   2454
Name: basin, dtype: int64
```

In [77]:

```
#Geographic location
df['subvillage'].value_counts()
```

Out[77]:

```
Majengo      494
Shuleni      492
Madukani     435
Unknown      371
Kati         366
...
Igodimwa     1
Jengemwanama 1
Buyoga A     1
Ilolangulu Busenda B 1
Bombambili 2 1
Name: subvillage, Length: 18568, dtype: int64
```

In [78]:

```
#geographic location
```

```
df['region'].value_counts()
```

Out[78]:

```
Iringa      5294
Mbeya       4639
Kilimanjaro 4379
Morogoro    4006
Shinyanga   3977
Arusha      3350
Kagera      3316
Kigoma      2816
Ruvuma      2640
Pwani       2635
Tanga       2547
Mwanza      2295
Dodoma      2201
Singida     2093
Mara        1969
Tabora      1959
Rukwa       1808
Mtwara      1730
Manyara     1583
Lindi       1546
Dar es Salaam 805
Name: region, dtype: int64
```

Both 'region_code' and 'district_code' are integer types. However, as codes, it's unlikely that there is any meaning to the order or progression of values. Therefore, we decided to treat them as categorical variables and converted them to strings.

In [79]:

```
#geographic location (coded)
df['region_code'].value_counts()
```

Out[79]:

```
11      5297
12      4639
3       4379
5       4040
17      3954
18      3324
2       3024
16      2816
10      2640
4       2513
19      2295
1       2201
13      2093
14      1979
20      1969
15      1808
6       1609
21      1583
80      1238
60      1025
90       917
7        805
99       423
9        390
24       326
8        300
40         1
Name: region_code, dtype: int64
```

In [80]:

```
#region code is type int, but best presented as a categorical feature
#convert to type string
```

```
df['region_code'] = df['region_code'].astype(str)
```

```
In [81]:
```

```
df['region_code']
```

```
Out[81]:
```

```
0      11
1      20
2      21
3      90
4      18
      ..
59395   3
59396  11
59397  12
59398   1
59399   5
Name: region_code, Length: 57588, dtype: object
```

```
In [82]:
```

```
#geographic location (coded)
df['district_code'].value_counts()
```

```
Out[82]:
```

```
1      11146
2      10909
3       9998
4       8996
5       4356
6       3586
7       3343
8       1043
30       995
33       874
53       745
43       505
13       391
23       293
63       195
62       109
60        63
0         23
80        12
67         6
Name: district_code, dtype: int64
```

```
In [83]:
```

```
#district code is type int, but best presented as a categorical feature
#convert to type string
df['district_code'] = df['district_code'].astype(str)
```

```
In [84]:
```

```
df['district_code']
```

```
Out[84]:
```

```
0         5
1         2
2         4
3        63
4         1
      ..
59395     5
59396     4
59397     7
59398     4
59399     2
```

Name: district_code, Length: 57588, dtype: object

Based on some research and information lookup, 'lga' seems to refer to cities or areas of cities.

In [85]:

```
#geographic location (city?)
df['lga'].value_counts()
```

Out[85]:

```
Njombe          2503
Arusha Rural    1252
Moshi Rural     1251
Rungwe          1106
Kilosa          1094
...
Moshi Urban      79
Kigoma Urban     71
Arusha Urban     63
Lindi Urban      21
Nyamagana        1
Name: lga, Length: 124, dtype: int64
```

In [86]:

```
#geographic location (ward?)
df['ward'].value_counts()
```

Out[86]:

```
Igosi           307
Imalinyi        252
Siha Kati       232
Mdandu          231
Nduruma         217
...
Nsemulwa        1
Ifinga          1
Kihangimahuka   1
Them            1
Mawenzi         1
Name: ward, Length: 2033, dtype: int64
```

The 'population' column has a large number of '0' values for the population around the pump.

In [87]:

```
#population around the well
df['population'].value_counts()
```

Out[87]:

```
0          19569
1           7025
200         1940
150         1892
250         1681
...
3241         1
1960         1
1685         1
2248         1
1439         1
Name: population, Length: 1049, dtype: int64
```

In [88]:

```
df['public_meeting'].value_counts()
```

Out[88]:

```
True      52713
False     4875
Name: public_meeting, dtype: int64
```

The 'recorded_by' column told us that the data was collected by a group called GeoData Consultants Ltd. This fact was useful for our data understanding but inessential to our modeling.

In [89]:

```
#feature unimportant for modeling
df['recorded_by'].value_counts()
```

Out[89]:

```
GeoData Consultants Ltd    57588
Name: recorded_by, dtype: int64
```

As mentioned earlier, there were numerous 0 values in 'construction_year'. 20,709 values represents about one-third of our total amount of data.

In [90]:

```
#how do we treat 0's in construction year?
#age as an ordinal encoded variable to properly treat 0's?
df['construction_year'].value_counts()
```

Out[90]:

```
0          18897
2010        2645
2008        2613
2009        2533
2000        2091
2007        1587
2006        1471
2003        1286
2011        1256
2004        1123
2012        1084
2002        1075
1978        1037
1995        1014
2005        1011
1999         979
1998         966
1990         954
1985         945
1980         811
1996         811
1984         779
1982         744
1994         738
1972         708
1974         676
1997         644
1992         640
1993         608
2001         540
1988         521
1983         488
1975         437
1986         434
1976         414
1970         411
1991         324
1989         316
1987         302
1981         238
1977         202
1979         192
1973         184
```

```
2013      176
1971      145
1960      102
1967       88
1963       85
1968       77
1969       59
1964       40
1962       30
1961       21
1965       19
1966       17
```

```
Name: construction_year, dtype: int64
```

```
In [91]:
```

```
df[df['construction_year'] != 0]['construction_year'].describe()
#construction years range from 1960 to 2013, with 20709 values of 0
```

```
Out[91]:
```

```
count      38691.000000
mean       1996.814686
std         12.472045
min        1960.000000
25%        1987.000000
50%        2000.000000
75%        2008.000000
max         2013.000000
Name: construction_year, dtype: float64
```

From the known construction years, we saw a range from 1960 to 2013. This gave us an idea of the vastness of this data set.

Similar Columns

The following section identifies groups of columns that all contain similar information. Within these groups, some columns are more granular than others. For modeling, we only wanted to use a single column for each group so that we wouldn't have multiple features all encapsulating essentially the same information.

```
In [92]:
```

```
df['extraction_type'].value_counts()
```

```
Out[92]:
```

```
gravity                26696
nira/tanira            7361
other                  6160
submersible            4688
swn 80                 3448
mono                   2817
india mark ii          2284
afridev                1659
ksb                    1358
other - rope pump       451
other - swn 81          229
windmill               117
india mark iii          91
cemo                   90
other - play pump       85
climax                 32
walimi                 20
other - mkulima/shinyanga 2
Name: extraction_type, dtype: int64
```

```
In [93]:
```

```
df['extraction_type_group'].value_counts()
```

Out[93]:

```
gravity          26696
nira/tanira      7361
other            6160
submersible      6046
swn 80           3448
mono             2817
india mark ii    2284
afridev          1659
rope pump        451
other handpump   336
other motorpump  122
wind-powered     117
india mark iii   91
Name: extraction_type_group, dtype: int64
```

In [94]:

```
df['extraction_type_class'].value_counts()
```

Out[94]:

```
gravity          26696
handpump         15179
other            6160
submersible      6046
motorpump        2939
rope pump         451
wind-powered     117
Name: extraction_type_class, dtype: int64
```

'extraction_type' is the most granular, and 'extraction_type_class' is the most broad.

In [95]:

```
#how does this differ from scheme_management
df['management'].value_counts()
```

Out[95]:

```
vwc              39746
wug              5556
water board      2932
wua              2533
private operator 1970
parastatal       1696
water authority   902
other            840
company          685
unknown          551
other - school    99
trust            78
Name: management, dtype: int64
```

In [96]:

```
df['scheme_management'].value_counts()
```

Out[96]:

```
VWC              36143
WUG              4249
None             3751
Water authority  3151
WUA              2882
Water Board      2747
Parastatal       1607
Private operator 1063
Company          1061
Other            765
SWC              97
```



```
Trust 72
Name: scheme_management, dtype: int64
```

In [97]:

```
df['management_group'].value_counts()
```

Out[97]:

```
user-group    50767
commercial    3635
parastatal    1696
other          939
unknown       551
Name: management_group, dtype: int64
```

'management' is the most granular, and 'management_group' is the most broad. Capitalization was also inconsistent between columns.

In [98]:

```
df['payment'].value_counts()
```

Out[98]:

```
never pay          24380
pay per bucket     8953
pay monthly        8229
unknown            7654
pay when scheme fails 3843
pay annually       3626
other              903
Name: payment, dtype: int64
```

In [99]:

```
df['payment_type'].value_counts()
```

Out[99]:

```
never pay    24380
per bucket   8953
monthly      8229
unknown      7654
on failure   3843
annually     3626
other        903
Name: payment_type, dtype: int64
```

The value counts for 'payment' and 'payment_type' aligned perfectly. They relayed the same information with differently named categories.

In [100]:

```
df['water_quality'].value_counts()
```

Out[100]:

```
soft          49431
salty         4772
unknown       1661
milky         803
coloured      479
salty abandoned 228
fluoride      199
fluoride abandoned 15
Name: water_quality, dtype: int64
```

In [101]:

```
df['quality_group'].value_counts()
```

```
Out[101]:
```

```
good          49431
salty         5000
unknown      1661
milky         803
colored       479
fluoride      214
Name: quality_group, dtype: int64
```

'water_quality' and 'quality_group' convey the same information with some slight differences in values. water_quality is more granular.

```
In [102]:
```

```
df['quantity'].value_counts()
```

```
Out[102]:
```

```
enough          32260
insufficient    14564
dry             5990
seasonal        4001
unknown         773
Name: quantity, dtype: int64
```

```
In [103]:
```

```
df['quantity_group'].value_counts()
```

```
Out[103]:
```

```
enough          32260
insufficient    14564
dry             5990
seasonal        4001
unknown         773
Name: quantity_group, dtype: int64
```

'quantity' and 'quantity_group' had identical value counts and category names.

```
In [104]:
```

```
df['source'].value_counts()
```

```
Out[104]:
```

```
spring          17006
shallow well    15499
machine dbh     10826
river           9612
rainwater harvesting 2218
hand dtw        873
dam             649
lake            639
other           202
unknown         64
Name: source, dtype: int64
```

```
In [105]:
```

```
df['source_type'].value_counts()
```

```
Out[105]:
```

```
spring          17006
shallow well    15499
borehole        11699
river/lake      10251
rainwater harvesting 2218
dam             649
other           266
```

```
Out[105]:  
Name: source_type, dtype: int64
```

```
In [106]:
```

```
df['source_class'].value_counts()
```

```
Out[106]:
```

```
groundwater    44204  
surface        13118  
unknown         266  
Name: source_class, dtype: int64
```

'source' is the most granular, and 'source_class' is the most broad.

```
In [107]:
```

```
df['waterpoint_type'].value_counts()
```

```
Out[107]:
```

```
communal standpipe    28375  
hand pump             16181  
other                 6167  
communal standpipe multiple    5959  
improved spring       783  
cattle trough         116  
dam                   7  
Name: waterpoint_type, dtype: int64
```

```
In [108]:
```

```
df['waterpoint_type_group'].value_counts()
```

```
Out[108]:
```

```
communal standpipe    34334  
hand pump             16181  
other                 6167  
improved spring       783  
cattle trough         116  
dam                   7  
Name: waterpoint_type_group, dtype: int64
```

waterpoint_type and waterpoint_type_group are almost identical except the communal standpipe multiple group in 'waterpoint_type' is included in 'communal standpipe' in 'waterpoint_type_group'

```
In [109]:
```

```
df['id'].value_counts()
```

```
Out[109]:
```

```
2047      1  
20959     1  
4759      1  
661       1  
2708      1  
..  
62836     1  
52595     1  
50546     1  
56689     1  
0         1  
Name: id, Length: 57588, dtype: int64
```

```
In [110]:
```

```
df['id_label'].value_counts()
```

```
Out[110]:
```

2047 1
20959 1
4759 1
661 1
2708 1
..
62836 1
52595 1
50546 1
56689 1
0 1
Name: id_label, Length: 57588, dtype: int64

We confirmed that no duplicate 'id' appeared in the dataset

In [111]:

```
df.head()
```

Out[111]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
0	69572	6000.0	2011-03-14	Roman	1390	Roman	34.938093	-9.856322	none	0	...	
1	8776	0.0	2013-03-06	Grumeti	1399	GRUMETI	34.698766	-2.147466	Zahanati	0	...	
2	34310	25.0	2013-02-25	Lottery Club	686	World vision	37.460664	-3.821329	Kwa Mahundi	0	...	
3	67743	0.0	2013-01-28	Unicef	263	UNICEF	38.486161	-11.155298	Zahanati Ya Nanyumbu	0	...	
4	19728	0.0	2011-07-13	Action In A	0	Artisan	31.130847	-1.825359	Shuleni	0	...	

5 rows x 42 columns



In [112]:

```
#FINAL NULLS CHECK
df.isna().sum()
```

Out[112]:

id 0
amount_tsh 0
date_recorded 0
funder 0
gps_height 0
installer 0
longitude 0
latitude 0
wpt_name 0
num_private 0
basin 0
subvillage 0
region 0
region_code 0
district_code 0
lga 0
ward 0
population 0
public_meeting 0
recorded_by 0
scheme_management 0
scheme_name 0
permit 0

```

construction_year      0
extraction_type        0
extraction_type_group  0
extraction_type_class  0
management             0
management_group       0
payment               0
payment_type           0
water_quality          0
quality_group          0
quantity              0
quantity_group         0
source                0
source_type            0
source_class           0
waterpoint_type        0
waterpoint_type_group  0
id_label              0
status_group          0
dtype: int64

```

Our cleaning removed all nulls in the dataset.

After exploring the columns, we adjusted some values that appeared as placeholders. We also identified columns that continued similar or identical information and determined how we would only implement one column from every similar group into our model.

In [113]:

```
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 57588 entries, 0 to 59399
Data columns (total 42 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   id                    57588 non-null  int64
 1   amount_tsh           57588 non-null  float64
 2   date_recorded        57588 non-null  object
 3   funder               57588 non-null  object
 4   gps_height           57588 non-null  int64
 5   installer            57588 non-null  object
 6   longitude            57588 non-null  float64
 7   latitude             57588 non-null  float64
 8   wpt_name             57588 non-null  object
 9   num_private          57588 non-null  int64
10   basin                57588 non-null  object
11   subvillage           57588 non-null  object
12   region               57588 non-null  object
13   region_code          57588 non-null  object
14   district_code        57588 non-null  object
15   lga                  57588 non-null  object
16   ward                 57588 non-null  object
17   population           57588 non-null  int64
18   public_meeting       57588 non-null  bool
19   recorded_by          57588 non-null  object
20   scheme_management    57588 non-null  object
21   scheme_name          57588 non-null  object
22   permit              57588 non-null  bool
23   construction_year    57588 non-null  int64
24   extraction_type       57588 non-null  object
25   extraction_type_group 57588 non-null  object
26   extraction_type_class 57588 non-null  object
27   management           57588 non-null  object
28   management_group     57588 non-null  object
29   payment              57588 non-null  object
30   payment_type         57588 non-null  object
31   water_quality        57588 non-null  object
32   quality_group        57588 non-null  object
33   quantity             57588 non-null  object
34   quantity_group       57588 non-null  object

```

```

35 source 57588 non-null object
36 source_type 57588 non-null object
37 source_class 57588 non-null object
38 waterpoint_type 57588 non-null object
39 waterpoint_type_group 57588 non-null object
40 id_label 57588 non-null int64
41 status_group 57588 non-null object
dtypes: bool(2), float64(3), int64(6), object(31)
memory usage: 18.1+ MB

```

In [114]:

```
df.head()
```

Out[114]:

	id	amount_tsh	date_recorded	funder	gps_height	installer	longitude	latitude	wpt_name	num_private	...	qua
0	69572	6000.0	2011-03-14	Roman	1390	Roman	34.938093	-9.856322	none	0	...	
1	8776	0.0	2013-03-06	Grumeti	1399	GRUMETI	34.698766	-2.147466	Zahanati	0	...	
2	34310	25.0	2013-02-25	Lottery Club	686	World vision	37.460664	-3.821329	Kwa Mahundi	0	...	
3	67743	0.0	2013-01-28	Unicef	263	UNICEF	38.486161	11.155298	Zahanati Ya Nanyumbu	0	...	
4	19728	0.0	2011-07-13	Action In A	0	Artisan	31.130847	-1.825359	Shuleni	0	...	

5 rows x 42 columns



We saved the dataframe containing all fo the cleaned data and the 'target_group' into a separate CSV file available in the data folder of this project's repository. This file was imported into our modeling notebook.

In [115]:

```

#SAVE CLEAN DATA OFF TO CSV FILE FOR IMPORT INTO OTHER NOTEBOOKS
#index= False
#df.to_csv('./data/water_well_train_clean.csv', index_label=False)

```

Please proceed to the modeling notebook.