In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

In [2]:

mining_data=pd.read_csv(r'Mining_Dataset.csv', decimal=',',parse_dates=["date"],infer_datet

In [3]:

mining_data.head()

Out[3]:

	date	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	
0	2017- 03-10 01:00:00	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	
1	2017- 03-10 01:00:00	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	
2	2017- 03-10 01:00:00	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	
3	2017- 03-10 01:00:00	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	
4	2017- 03-10 01:00:00	55.2	16.98	3033.69	558.167	400.254	10.0697	1.74	250.203	252.136	

5 rows × 24 columns

In [4]:

mining_data.head(15)

Out[4]:

	date	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	
0	2017- 03-10 01:00:00	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	
1	2017- 03-10 01:00:00	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	
2	2017- 03-10 01:00:00	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	
3	2017- 03-10 01:00:00	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	
4	2017- 03-10 01:00:00	55.2	16.98	3033.69	558.167	400.254	10.0697	1.74	250.203	252.136	
5	2017- 03-10 01:00:00	55.2	16.98	3079.10	564.697	396.533	10.0705	1.74	250.730	248.906	
6	2017- 03-10 01:00:00	55.2	16.98	3127.79	566.467	392.900	10.0713	1.74	250.313	252.202	
7	2017- 03-10 01:00:00	55.2	16.98	3152.93	558.777	397.002	10.0722	1.74	249.895	253.630	
8	2017- 03-10 01:00:00	55.2	16.98	3147.27	556.030	394.307	10.0730	1.74	250.137	251.104	
9	2017- 03-10 01:00:00	55.2	16.98	3142.58	565.857	393.105	10.0738	1.74	249.653	252.202	
10	2017- 03-10 01:00:00	55.2	16.98	3148.05	561.951	396.533	10.0746	1.74	249.236	250.818	
11	2017- 03-10 01:00:00	55.2	16.98	3150.39	558.472	397.852	10.0755	1.74	249.170	249.829	
12	2017- 03-10 01:00:00	55.2	16.98	3280.27	564.026	393.545	10.0763	1.74	249.016	249.829	
13	2017- 03-10 01:00:00	55.2	16.98	3411.13	567.261	394.160	10.0771	1.74	249.258	250.137	
14	2017- 03-10 01:00:00	55.2	16.98	3447.46	561.646	392.549	10.0779	1.74	249.390	251.191	

15 rows × 24 columns

In [5]:

mining_data.describe()

Out[5]:

	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp
count	737453.000000	737453.000000	737453.000000	737453.000000	737453.000000	737453.0000
mean	56.294739	14.651716	2869.140569	488.144697	397.578372	9.7676
std	5.157744	6.807439	1215.203734	91.230534	9.699785	0.3870
min	42.740000	1.310000	0.002026	241.669000	376.249000	8.7530
25%	52.670000	8.940000	2076.320000	431.796000	394.264000	9.5270
50%	56.080000	13.850000	3018.430000	504.393000	399.249000	9.798
75%	59.720000	19.600000	3727.730000	553.257000	402.968000	10.0380
max	65.780000	33.400000	6300.230000	739.538000	418.641000	10.808
8 rows	× 23 columns					
4						•

In [6]:

mining_data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 737453 entries, 0 to 737452
Data columns (total 24 columns):
```

Column Non-Null Count Dtype ---------_ _ _ _ _ 0 date 737453 non-null datetime64[ns] 1 % Iron Feed 737453 non-null float64 2 % Silica Feed float64 737453 non-null 3 Starch Flow 737453 non-null float64 Amina Flow 4 737453 non-null float64 5 Ore Pulp Flow 737453 non-null float64 Ore Pulp pH 6 737453 non-null float64 7 Ore Pulp Density 737453 non-null float64 8 Flotation Column 01 Air Flow 737453 non-null float64 Flotation Column 02 Air Flow 9 737453 non-null float64 Flotation Column 03 Air Flow 737453 non-null float64 10 11 Flotation Column 04 Air Flow 737453 non-null float64 Flotation Column 05 Air Flow 12 737453 non-null float64 Flotation Column 06 Air Flow 13 737453 non-null float64 Flotation Column 07 Air Flow 737453 non-null float64 Flotation Column 01 Level 737453 non-null float64 Flotation Column 02 Level 737453 non-null float64 Flotation Column 03 Level 17 737453 non-null float64 18 Flotation Column 04 Level 737453 non-null float64 Flotation Column 05 Level 737453 non-null float64 19 Flotation Column 06 Level 737453 non-null float64

737453 non-null float64

737453 non-null float64

737453 non-null float64

dtypes: datetime64[ns](1), float64(23)

Flotation Column 07 Level

% Iron Concentrate

% Silica Concentrate

memory usage: 135.0 MB

In [7]:

```
for cols in mining_data.columns.tolist()[1:]:
    df = mining_data.loc[mining_data[cols] > 0]
df.info()
```

<class 'pandas.core.frame.DataFrame'> Int64Index: 737453 entries, 0 to 737452 Data columns (total 24 columns):

#	Column	Non-Null Count	Dtype
0	date	737453 non-null	<pre>datetime64[ns]</pre>
1	% Iron Feed	737453 non-null	float64
2	% Silica Feed	737453 non-null	float64
3	Starch Flow	737453 non-null	float64
4	Amina Flow	737453 non-null	float64
5	Ore Pulp Flow	737453 non-null	float64
6	Ore Pulp pH	737453 non-null	float64
7	Ore Pulp Density	737453 non-null	float64
8	Flotation Column 01 Air Flow	737453 non-null	float64
9	Flotation Column 02 Air Flow	737453 non-null	float64
10	Flotation Column 03 Air Flow	737453 non-null	float64
11	Flotation Column 04 Air Flow	737453 non-null	float64
12	Flotation Column 05 Air Flow	737453 non-null	float64
13	Flotation Column 06 Air Flow	737453 non-null	float64
14	Flotation Column 07 Air Flow	737453 non-null	float64
15	Flotation Column 01 Level	737453 non-null	float64
16	Flotation Column 02 Level	737453 non-null	float64
17	Flotation Column 03 Level		float64
18	Flotation Column 04 Level	737453 non-null	float64
19	Flotation Column 05 Level	737453 non-null	float64
20	Flotation Column 06 Level	737453 non-null	float64
	Flotation Column 07 Level		
22	% Iron Concentrate	737453 non-null	float64
23	% Silica Concentrate	737453 non-null	float64
dtyp	es: datetime64[ns](1), float64	(23)	

dtypes: datetime64[ns](1), float64(23)

memory usage: 140.7 MB

In [8]:

```
mining_data.isnull().any()
```

Out[8]:

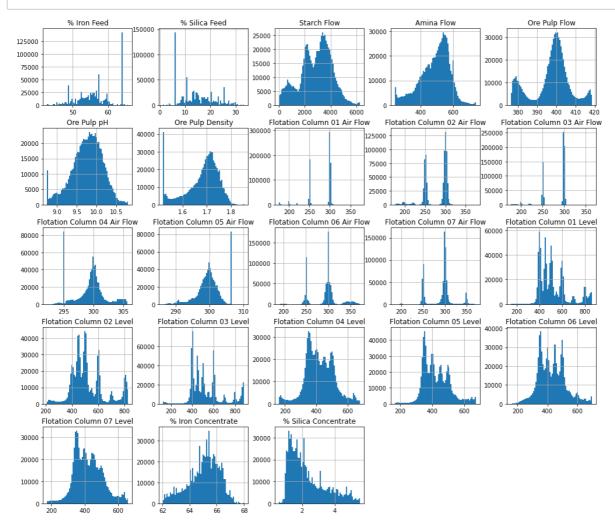
```
date
                                 False
% Iron Feed
                                 False
% Silica Feed
                                 False
Starch Flow
                                 False
Amina Flow
                                 False
Ore Pulp Flow
                                 False
Ore Pulp pH
                                 False
Ore Pulp Density
                                 False
Flotation Column 01 Air Flow
                                 False
Flotation Column 02 Air Flow
                                 False
Flotation Column 03 Air Flow
                                 False
Flotation Column 04 Air Flow
                                False
Flotation Column 05 Air Flow
                                 False
Flotation Column 06 Air Flow
                                 False
Flotation Column 07 Air Flow
                                 False
Flotation Column 01 Level
                                 False
Flotation Column 02 Level
                                 False
Flotation Column 03 Level
                                 False
Flotation Column 04 Level
                                 False
Flotation Column 05 Level
                                 False
Flotation Column 06 Level
                                 False
Flotation Column 07 Level
                                 False
% Iron Concentrate
                                False
% Silica Concentrate
                                False
dtype: bool
```

In [9]:

```
df=df.set_index('date')
```

In [11]:

```
import matplotlib.pyplot as plt
from matplotlib import style
df.hist(bins = 70, figsize = (17,15))
plt.show()
plt.suptitle('figure title', color='green')
```



Out[11]:

Text(0.5, 0.98, 'figure title')

<Figure size 432x288 with 0 Axes>

In [12]:

df.head()

Out[12]:

	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	Flotatic Colun 03 A Flo
date										
2017-03- 10 01:00:00	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	250.5
2017-03- 10 01:00:00	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	250.86
2017-03- 10 01:00:00	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	250.3 ⁻
2017-03- 10 01:00:00	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	250.04
2017-03- 10 01:00:00	55.2	16.98	3033.69	558.167	400.254	10.0697	1.74	250.203	252.136	249.89

5 rows × 23 columns

In [13]:

```
df_copy=df
from scipy import stats
df_copy=df[(np.abs(stats.zscore(df)) < 2).all(axis=1)]
df_copy.info()</pre>
```

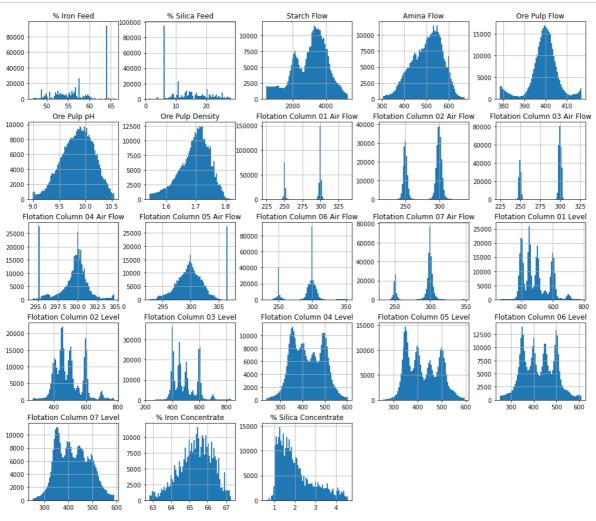
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 346475 entries, 2017-03-10 01:00:00 to 2017-09-09 22:00:00
Data columns (total 23 columns):

#	Column	Non-Null Count	Dtype
0	% Iron Feed	346475 non-null	float64
1	% Silica Feed	346475 non-null	float64
2	Starch Flow	346475 non-null	float64
3	Amina Flow	346475 non-null	float64
4	Ore Pulp Flow	346475 non-null	float64
5	Ore Pulp pH	346475 non-null	float64
6	Ore Pulp Density	346475 non-null	float64
7	Flotation Column 01 Air Flow	346475 non-null	float64
8	Flotation Column 02 Air Flow	346475 non-null	float64
9	Flotation Column 03 Air Flow	346475 non-null	float64
10	Flotation Column 04 Air Flow	346475 non-null	float64
11	Flotation Column 05 Air Flow	346475 non-null	float64
12	Flotation Column 06 Air Flow	346475 non-null	float64
13	Flotation Column 07 Air Flow	346475 non-null	float64
14	Flotation Column 01 Level	346475 non-null	float64
15	Flotation Column 02 Level	346475 non-null	float64
16	Flotation Column 03 Level	346475 non-null	float64
17	Flotation Column 04 Level	346475 non-null	float64
18	Flotation Column 05 Level	346475 non-null	float64
19	Flotation Column 06 Level	346475 non-null	float64
20	Flotation Column 07 Level	346475 non-null	float64
21	% Iron Concentrate	346475 non-null	float64
22	% Silica Concentrate	346475 non-null	float64
	63		

dtypes: float64(23)
memory usage: 63.4 MB

In [14]:

```
df_copy.hist(bins = 70, figsize = (17,15))
plt.show()
plt.suptitle('figure title', color='w')
df_copy.info()
```



DatetimeIndex: 346475 entries, 2017-03-10 01:00:00 to 2017-09-09 22:00:00 Data columns (total 23 columns):

#	Column	Non-Null Count	Dtype
0		346475 non-null	
1	% Silica Feed	346475 non-null	float64
2	Starch Flow	346475 non-null	float64
3	Amina Flow	346475 non-null	float64
4	Ore Pulp Flow	346475 non-null	float64
5	Ore Pulp pH	346475 non-null	float64
6	Ore Pulp Density	346475 non-null	float64
7	Flotation Column 01 Air Flow	346475 non-null	float64
8	Flotation Column 02 Air Flow	346475 non-null	float64
9	Flotation Column 03 Air Flow	346475 non-null	float64
10	Flotation Column 04 Air Flow	346475 non-null	float64
11	Flotation Column 05 Air Flow	346475 non-null	float64
12	Flotation Column 06 Air Flow	346475 non-null	float64
13	Flotation Column 07 Air Flow	346475 non-null	float64
14	Flotation Column 01 Level	346475 non-null	float64
15	Flotation Column 02 Level	346475 non-null	float64
16	Flotation Column 03 Level	346475 non-null	float64
17	Flotation Column 04 Level	346475 non-null	float64
18	Flotation Column 05 Level	346475 non-null	float64
19	Flotation Column 06 Level	346475 non-null	float64
20	Flotation Column 07 Level	346475 non-null	float64
21	% Iron Concentrate	346475 non-null	float64
22	% Silica Concentrate	346475 non-null	float64

dtypes: float64(23)
memory usage: 63.4 MB

<Figure size 432x288 with 0 Axes>

In [15]:

```
df_copy.drop(columns=['% Iron Concentrate'],inplace=True)
df_copy.head()
```

C:\Users\akhil\anaconda3\lib\site-packages\pandas\core\frame.py:4308: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

return super().drop(

Out[15]:

	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	Flotatic Colun 03 A Flo
date										
2017-03- 10 01:00:00	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	250.5
2017-03- 10 01:00:00	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	250.86
2017-03- 10 01:00:00	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	250.3 ⁻
2017-03- 10 01:00:00	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	250.0 ₄
2017-03- 10 01:00:00	55.2	16.98	3033.69	558.167	400.254	10.0697	1.74	250.203	252.136	249.89

5 rows × 22 columns

In [16]:

mining_data.head()

Out[16]:

	date	% Iron Feed	% Silica Feed	Starch Flow	Amina Flow	Ore Pulp Flow	Ore Pulp pH	Ore Pulp Density	Flotation Column 01 Air Flow	Flotation Column 02 Air Flow	
0	2017- 03-10 01:00:00	55.2	16.98	3019.53	557.434	395.713	10.0664	1.74	249.214	253.235	
1	2017- 03-10 01:00:00	55.2	16.98	3024.41	563.965	397.383	10.0672	1.74	249.719	250.532	
2	2017- 03-10 01:00:00	55.2	16.98	3043.46	568.054	399.668	10.0680	1.74	249.741	247.874	
3	2017- 03-10 01:00:00	55.2	16.98	3047.36	568.665	397.939	10.0689	1.74	249.917	254.487	
4	2017- 03-10 01:00:00	55.2	16.98	3033.69	558.167	400.254	10.0697	1.74	250.203	252.136	

5 rows × 24 columns

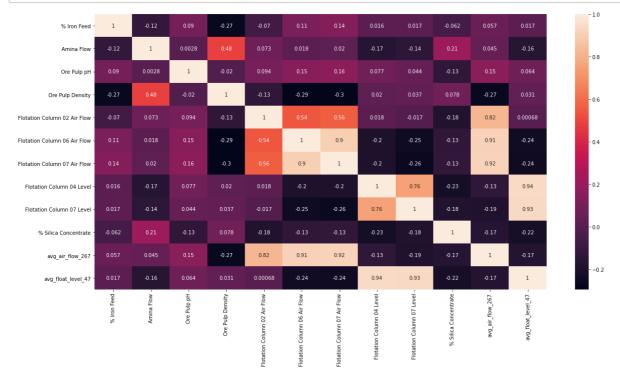
In [17]:

```
import seaborn as sns
ml_mining_data=df_copy
plt.figure(figsize=(20, 20))
p = sns.heatmap(ml_mining_data.corr(), annot=True);
```

% Iron Feed -	. 1	-0.97	0.092	-0.12	0.12	0.09	-0.27	0.09	-0.07	0.091	-0.0056	0.055	0.11	0.14	0.052	0.041	0.03	0.016	0.017	0.014	0.017	-0.062		-1.00
% Silica Feed -	-0.97	1	-0.092		-0.12			-0.044			0.044				-0.044							0.052		
Starch Flow	0.092	-0.092	1	0.063	0.042		-0.0092		0.042	0.043		-0.049			-0.19	-0.18	-0.18					-0.053		
Amina Flow -	-0.12		0.063	1	0.21		0.48			0.042		-0.045			-0.27	-0.25	-0.28	-0.17	-0.18	-0.13	-0.14	0.21		- 0.75
Ore Pulp Flow	0.12	-0.12	0.042	0.21	1		0.0071	-0.11	-0.085	-0.11	-0.064	0.042	-0.042				-0.084					0.053		
Ore Pulp pH -					0.057	1	-0.02															-0.13		- 0.50
Ore Pulp Density -	-0.27		-0.0092	0.48	0.0071	-0.02	1	-0.21	-0.13	-0.22	-0.25		-0.29	-0.3						0.041		0.078		
Flotation Column 01 Air Flow -		-0.044		0.029	-0.11		-0.21	1	0.77	0.96	0.44	-0.16	0.65	0.65	-0.26	-0.23	-0.21			0.0066		-0.24		
Flotation Column 02 Air Flow -			0.042				-0.13	0.77	1	0.8		-0.12		0.56	-0.29	-0.29	-0.25		0.0046	-0.0043		-0.18		- 0.25
Flotation Column 03 Air Flow				0.042	-0.11		-0.22	0.96	0.8	1		-0.16	0.67	0.67	-0.26	-0.23	-0.22					-0.24		
Flotation Column 04 Air Flow -	-0.0056	0.044					-0.25	0.44	0.37	0.46	1	-0.28		0.57		-0.18	-0.14		-0.22	-0.24	-0.24	0.0042		
Flotation Column 05 Air Flow			-0.049	-0.045	0.042			-0.16	-0.12	-0.16	-0.28	1	-0.22	-0.22								0.02		- 0.00
Flotation Column 06 Air Flow -					-0.042		-0.29	0.65	0.54	0.67	0.55	-0.22	1	0.9	-0.36	-0.32	-0.33	-0.2	-0.22	-0.22	-0.25	-0.13		
Flotation Column 07 Air Flow -		-0.086					-0.3	0.65	0.56	0.67	0.57	-0.22	0.9	1	-0.37	-0.34	-0.33		-0.22	-0.24	-0.26	-0.13		
Flotation Column 01 Level		-0.044	-0.19	-0.27				-0.26	-0.29	-0.26	-0.2	0.14	-0.36	-0.37	1	0.79	0.8	0.35	0.39	0.33	0.35	-0.064		0.25
Flotation Column 02 Level -			-0.18	-0.25				-0.23	-0.29	-0.23	-0.18		-0.32	-0.34	0.79	1	0.75					-0.059		
Flotation Column 03 Level			-0.18	-0.28				-0.21	-0.25	-0.22	-0.14		-0.33	-0.33	0.8	0.75	1					-0.06		0.50
Flotation Column 04 Level -	0.016			-0.17										-0.2	0.35	0.31	0.33	1	0.87	0.67	0.76	-0.23		0.50
Flotation Column 05 Level -				-0.18					0.0046		-0.22		-0.22	-0.22				0.87	1	0.71	0.81	-0.23		
Flotation Column 06 Level -				-0.13					-0.0043		-0.24		-0.22	-0.24				0.67	0.71	1	0.75	-0.17		0.75
Flotation Column 07 Level -	0.017			-0.14		0.044					-0.24		-0.25	-0.26				0.76	0.81	0.75	1	-0.18		
% Silica Concentrate -	-0.062					-0.13		-0.24	-0.18	-0.24	0.0042			-0.13	-0.064		-0.06	-0.23	-0.23	-0.17	-0.18	1		
	% Iron Feed -	% Silica Feed -	Starch Flow -	Amina Flow -	Ore Pulp Flow	Ore Pulp pH -	Ore Pulp Density -	otation Column 01 Air Flow -	otation Column 02 Air Flow -	otation Column 03 Air Flow -	otation Column 04 Air Flow -	otation Column 05 Air Flow -	otation Column 06 Air Flow -	otation Column 07 Air Flow -	Flotation Column 01 Level -	Flotation Column 02 Level -	Flotation Column 03 Level -	Flotation Column 04 Level -	Flotation Column 05 Level -	Flotation Column 06 Level -	Flotation Column 07 Level -	% Silica Concentrate -		

In [18]:

In [19]:



In [20]:

- 0.8

0.6

- 0.4

0.2

0.0

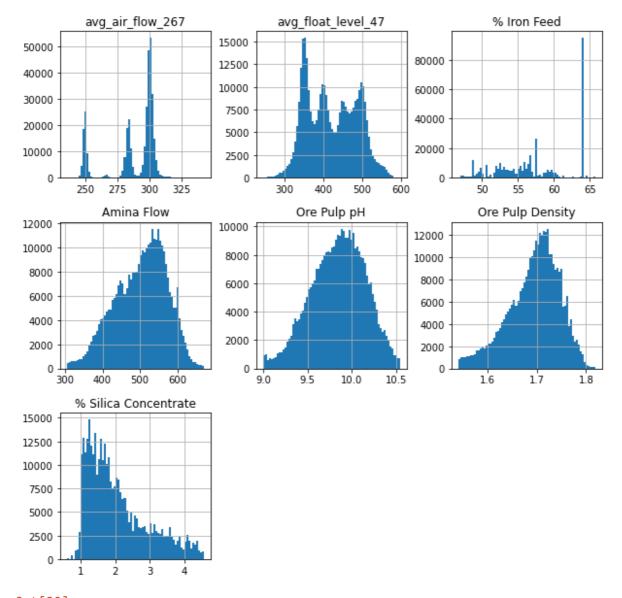
<class 'pandas.core.frame.DataFrame'>

DatetimeIndex: 346475 entries, 2017-03-10 01:00:00 to 2017-09-09 22:00:00 Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	avg_air_flow_267	346475 non-null	float64
1	avg_float_level_47	346475 non-null	float64
2	% Iron Feed	346475 non-null	float64
3	Amina Flow	346475 non-null	float64
4	Ore Pulp pH	346475 non-null	float64
5	Ore Pulp Density	346475 non-null	float64
6	% Silica Concentrate	346475 non-null	float64

dtypes: float64(7)
memory usage: 21.1 MB





Out[20]:

Text(0.5, 0.98, 'figure title')
<Figure size 432x288 with 0 Axes>

In []:

In [21]:

```
import pandas as pd
import numpy as np
```

In [23]:

```
dataset=pd.read_csv('Enhanced_Mining_dataset.csv',sep=',')
dataset=dataset.set_index('date')
dataset.head()
```

Out[23]:

	avg_air_flow_267	avg_float_level_47	% Iron Feed	Amina Flow	Ore Pulp pH	Ore Pulp Density	% Silica Concentrate
date							
2017-03- 10 01:00:00	251.448000	483.4510	55.2	557.434	10.0664	1.74	1.31
2017-03- 10 01:00:00	249.887667	473.0805	55.2	563.965	10.0672	1.74	1.31
2017-03- 10 01:00:00	249.096667	454.1275	55.2	568.054	10.0680	1.74	1.31
2017-03- 10 01:00:00	252.018667	436.9395	55.2	568.665	10.0689	1.74	1.31
2017-03- 10 01:00:00	250.349000	439.6745	55.2	558.167	10.0697	1.74	1.31

In [24]:

dataset.info()

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

Index: 346475 entries, 2017-03-10 01:00:00 to 2017-09-09 22:00:00

Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	avg_air_flow_267	346475 non-null	float64
1	avg_float_level_47	346475 non-null	float64
2	% Iron Feed	346475 non-null	float64
3	Amina Flow	346475 non-null	float64
4	Ore Pulp pH	346475 non-null	float64
5	Ore Pulp Density	346475 non-null	float64
6	% Silica Concentrate	346475 non-null	float64

dtypes: float64(7)
memory usage: 21.1+ MB

In [25]:

dataset.head()

Out[25]:

	avg_air_flow_267	avg_float_level_47	% Iron Feed	Amina Flow	Ore Pulp pH	Ore Pulp Density	% Silica Concentrate
date							
2017-03- 10 01:00:00	251.448000	483.4510	55.2	557.434	10.0664	1.74	1.31
2017-03- 10 01:00:00	249.887667	473.0805	55.2	563.965	10.0672	1.74	1.31
2017-03- 10 01:00:00	249.096667	454.1275	55.2	568.054	10.0680	1.74	1.31
2017-03- 10 01:00:00	252.018667	436.9395	55.2	568.665	10.0689	1.74	1.31
2017-03- 10 01:00:00	250.349000	439.6745	55.2	558.167	10.0697	1.74	1.31

In [26]:

```
x=dataset.iloc[:,:-1].values
y=dataset.iloc[:,-1].values
```

In [27]:

x

Out[27]:

```
, 483.451
array([[251.448
                                     55.2
                                                 , 557.434
         10.0664
                        1.74
                                  ],
                                     55.2
       [249.88766667, 473.0805
                                                 , 563.965
         10.0672
                      1.74
                                  ],
       [249.09666667, 454.1275
                                     55.2
                                                 , 568.054
         10.068
                        1.74
                                  ],
       [321.91983333, 444.47
                                     49.75
                                                 , 501.668
          9.75384
                        1.71849
                                  ],
                    , 434.3475
                                     49.75
                                                 , 501.514
       [320.522
                        1.71701
          9.74102
                                  ],
       [319.41466667, 389.8735
                                     49.75
                                                 , 483.383
          9.73784
                        1.7091
                                  ]])
```

```
In [28]:
У
Out[28]:
array([1.31, 1.31, 1.31, ..., 1.96, 1.96, 1.96])
In [29]:
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
In [30]:
from sklearn.ensemble import RandomForestRegressor
In [34]:
from sklearn.ensemble import RandomForestRegressor
model=RandomForestRegressor(n_estimators=42,criterion='mse')
model.fit(x_train,y_train)
Out[34]:
RandomForestRegressor(n_estimators=42)
In [35]:
y_pred=model.predict(x_test)
y_pred
Out[35]:
                              , 1.87309524, ..., 3.33
array([1.24952381, 1.23
                                                           , 2.20428571,
       2.8502381 ])
In [36]:
from sklearn.metrics import r2_score
r2_score(y_test,y_pred)
Out[36]:
0.9157556606287122
In [39]:
import pickle
pickle.dump(model,open('mining (1).pkl','wb'))
In [40]:
model.predict([[251.448000,483.4510,55.20,557.434,10.06640,1.74000]])
Out[40]:
array([1.29857143])
```

In []:			