

In [31]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from seaborn import clustermap
from sklearn.preprocessing import MinMaxScaler
%matplotlib notebook
```

In [32]:

```
cmm = pd.read_excel('CMMData.xlsx')
cmm
```

Out[32]:

	Unnamed: 0	Unnamed: 1	Unnamed: 2	Unnamed: 3	ID @100	ID @55	38 dia @200	42 dia @140	42 dia @80	Base angle F	...	162mm taper F	162mm
0	NaN	NaN	NaN	NaN	30.000	30.000	38.000	42.000	42.000	70.000	...	162.000	162.000
1	NaN	NaN	NaN	NaN	0.300	0.300	0.400	0.400	0.400	1.000	...	1.000	1.000
2	NaN	NaN	NaN	Part ID	-0.300	-0.300	-0.400	-0.400	-0.400	-1.000	...	-1.000	-1.000
3	31/10/2017	14:52:40	57.0	1	29.850	29.550	38.040	41.970	42.156	70.262	...	161.584	161.584
4	31/10/2017	15:10:33	58.0	2	29.853	29.564	38.036	41.963	42.154	70.181	...	161.672	161.672
...
79	06/11/2017	09:28:58	133.0	77	29.855	29.565	38.023	41.958	42.145	69.780	...	161.568	161.568
80	06/11/2017	09:44:34	134.0	78	29.864	29.561	38.019	41.939	42.141	70.044	...	161.558	161.558
81	06/11/2017	10:00:42	135.0	79	29.861	29.568	38.014	41.964	42.145	69.914	...	161.552	161.552

In [33]:

```
# check cmm datasets
cmm.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 84 entries, 0 to 83
Data columns (total 22 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Unnamed: 0            81 non-null    object
1   Unnamed: 1            81 non-null    object
2   Unnamed: 2            81 non-null    float64
3   Unnamed: 3            82 non-null    object
4   ID @100               84 non-null    float64
5   ID @55               84 non-null    float64
6   38 dia @200          84 non-null    float64
7   42 dia @140          84 non-null    float64
8   42 dia @80           84 non-null    float64
9   Base angle F          84 non-null    float64
10  Base angle BR         84 non-null    float64
11  Base angle BL         84 non-null    float64
12  162mm taper F         84 non-null    float64
13  162mm taper BR        84 non-null    float64
14  162mm taper BL        84 non-null    float64
15  40.5mm taper F        84 non-null    float64
16  40.5mm taper BR       84 non-null    float64
17  40.5mm taper BL       84 non-null    float64
18  Top1                  84 non-null    float64
19  Top2                  84 non-null    float64
20  Top3                  84 non-null    float64
21  Top4                  84 non-null    float64
dtypes: float64(19), object(3)
memory usage: 14.6+ KB
```

In [34]:

```
# check null values
cmm.isnull().sum()
# from table below, all values are 0. We can see that there are no NA values in cmm
```

Out[34]:

Unnamed: 0	3
Unnamed: 1	3
Unnamed: 2	3
Unnamed: 3	2
ID @100	0
ID @55	0
38 dia @200	0
42 dia @140	0
42 dia @80	0
Base angle F	0
Base angle BR	0
Base angle BL	0
162mm taper F	0
162mm taper BR	0
162mm taper BL	0
40.5mm taper F	0
40.5mm taper BR	0
40.5mm taper BL	0
Top1	0
Top2	0
Top3	0
Top4	0

dtype: int64

In [35]:

```
# rename features
cmm.rename(columns={'Unnamed: 3': 'Measurement'}, inplace=True)
cmm.drop(columns=['Unnamed: 0'], inplace=True)
cmm.drop(columns=['Unnamed: 1'], inplace=True)
cmm.drop(columns=['Unnamed: 2'], inplace=True)
cmm.iloc[0,0] = 'Nominal value'
cmm.iloc[1,0] = 'Upper error'
cmm.iloc[2,0] = 'Lower error'
cmm
```

Out[35]:

	Measurement	ID @100	ID @55	38 dia @200	42 dia @140	42 dia @80	Base angle F	Base angle BR	Base angle BL	162mm taper F	162m tap F
0	Nominal value	30.000	30.000	38.000	42.000	42.000	70.000	70.000	70.000	162.000	162.0
1	Upper error	0.300	0.300	0.400	0.400	0.400	1.000	1.000	1.000	1.000	1.0
2	Lower error	-0.300	-0.300	-0.400	-0.400	-0.400	-1.000	-1.000	-1.000	-1.000	-1.0
3	1	29.850	29.550	38.040	41.970	42.156	70.262	69.925	69.862	161.584	161.6
4	2	29.853	29.564	38.036	41.963	42.154	70.181	70.056	70.390	161.672	161.5
...
79	77	29.855	29.565	38.023	41.958	42.145	69.780	69.958	69.817	161.568	161.5
80	78	29.864	29.561	38.019	41.939	42.141	70.044	69.606	69.864	161.558	161.6
81	79	29.861	29.568	38.014	41.964	42.145	69.914	69.731	69.982	161.552	161.6
82	80	29.860	29.572	38.016	41.937	42.132	69.873	69.682	69.927	161.741	161.5
83	81	29.864	29.566	38.001	41.937	42.122	69.667	69.663	69.952	161.625	161.4

84 rows × 19 columns

As each measurement has different standard, I will transfer the nominal value to the same standard which is 1. Therefore, each measurement values are around 1 with different variance.

for loop, calculate the nominal values. For each column, I transfer all nominal value to 1 in order to get the same nominal values. To do that, I divide all values with their corresponding column's nominal values. By doing so, all measurement will have the same standard value as 1, and each measurement value differs around 1.

In [36]:

```
# For loop, standardize nominal values into 1.
for i in range(1, len(cmm.columns)):
    cmm.iloc[:, i] = cmm.iloc[:, i] / cmm.iloc[0,i]
cmm
```

	ID @100	ID @55	38 dia @200	42 dia @140	42 dia @80	Base angle F	Base angle BR	Base angle BL	162mm taper F	162mm taper BR	16 tap
	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
	0.010000	0.010000	0.010526	0.009524	0.009524	0.014286	0.014286	0.014286	0.006173	0.006173	0.006173
	-0.010000	-0.010000	-0.010526	-0.009524	-0.009524	-0.014286	-0.014286	-0.014286	-0.006173	-0.006173	-0.006173
	0.995000	0.985000	1.001053	0.999286	1.003714	1.003743	0.998929	0.998029	0.997432	0.998019	0.998019
	0.995100	0.985467	1.000947	0.999119	1.003667	1.002586	1.000800	1.005571	0.997975	0.997315	0.997315

	0.995167	0.985500	1.000605	0.999000	1.003452	0.996857	0.999400	0.997386	0.997333	0.997241	0.997241
	0.995467	0.985367	1.000500	0.998548	1.003357	1.000629	0.994371	0.998057	0.997272	0.997562	0.997562
	0.995367	0.985600	1.000368	0.999143	1.003452	0.998771	0.996157	0.999743	0.997235	0.997617	0.997617

In [37]:

```
# Build measurement standard data
cmm_measure = cmm.iloc[:3]
cmm_measure
```

Out[37]:

	Measurement	ID @100	ID @55	38 dia @200	42 dia @140	42 dia @80	Base angle F	Base angle BR	Base angle BL
0	Nominal value	1.00	1.00	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
1	Upper error	0.01	0.01	0.010526	0.009524	0.009524	0.014286	0.014286	0.014286
2	Lower error	-0.01	-0.01	-0.010526	-0.009524	-0.009524	-0.014286	-0.014286	-0.014286

In [38]:

```
# Build measure values for each part ID
cmm_part = cmm.iloc[3:].rename(columns={'Measurement': 'Part_ID'})
# Set ID index
cmm_part.set_index('Part_ID', inplace=True)
cmm_part
```

Out[38]:

	ID @100	ID @55	38 dia @200	42 dia @140	42 dia @80	Base angle F	Base angle BR	Base angle BL	162r tape
Part_ID									
1	0.995000	0.985000	1.001053	0.999286	1.003714	1.003743	0.998929	0.998029	0.9974
2	0.995100	0.985467	1.000947	0.999119	1.003667	1.002586	1.000800	1.005571	0.9979
3	0.995133	0.985267	1.000632	0.998929	1.003357	1.000386	0.998629	0.996871	0.9977
4	0.995167	0.985433	1.000447	0.998929	1.003238	0.996700	0.995786	0.996971	0.9984
5	0.995333	0.985700	1.000184	0.998643	1.003214	0.999386	0.996300	0.999857	0.9972
...
77	0.995167	0.985500	1.000605	0.999000	1.003452	0.996857	0.999400	0.997386	0.9973
78	0.995467	0.985367	1.000500	0.998548	1.003357	1.000629	0.994371	0.998057	0.9972
79	0.995367	0.985600	1.000368	0.999143	1.003452	0.998771	0.996157	0.999743	0.9972
80	0.995333	0.985733	1.000421	0.998500	1.003143	0.998186	0.995457	0.998957	0.9984
81	0.995467	0.985533	1.000026	0.998500	1.002905	0.995243	0.995186	0.999314	0.9976

81 rows × 18 columns

To calculate the total error for each part ID. As errors in 18 measures differs in positive and negative values, simply sum the error will cause positive error to cancel out the negative ones. In addition, each measurements' error difference is very small. Thus, I will square each error in 18 measurements to magnify errors, and then take the sum as the total error for each part ID.

In [39]:

```
cmm_part['total_error'] = 0
for i in range(0,len(cmm_part.columns)-1):
    # calculate the square for each error, minus nominal values 1
    cmm_part.iloc[:,i] = np.square(cmm_part.iloc[:,i]-1)
    cmm_part['total_error'] = cmm_part.iloc[:,i] + cmm_part['total_error'] # calculate total error

cmm_part
```

Out[39]:

	ID @100	ID @55	38 dia @200	42 dia @140	42 dia @80	Base angle F	Base angle BR	Base angle BL	162mm taper F	162mm taper BL
Part_ID										
1	0.000025	0.000225	1.108033e-06	5.102041e-07	0.000014	1.400898e-05	1.147959e-06	3.886531e-06	0.000007	0.000007
2	0.000024	0.000211	8.975069e-07	7.760771e-07	0.000013	6.685918e-06	6.400000e-07	3.104082e-05	0.000004	0.000004
3	0.000024	0.000217	3.988920e-07	1.147959e-06	0.000011	1.487755e-07	1.880816e-06	9.787959e-06	0.000005	0.000005
4	0.000023	0.000212	2.001385e-07	1.147959e-06	0.000010	1.089000e-05	1.776020e-05	9.172245e-06	0.000002	0.000002
5	0.000022	0.000204	3.393352e-08	1.841837e-06	0.000010	3.773469e-07	1.369000e-05	2.040816e-08	0.000007	0.000007
...

From the table above, we can see the squared error for each part and the total error for each Part ID.

In [40]:

```
# Rank total errors
cmm_part.sort_values(by='total_error', inplace=True, ascending=False)
cmm_part[['total_error']].reset_index()
```

Out[40]:

	Part_ID	total_error
0	23	0.017028
1	28	0.016768
2	32	0.016551
3	27	0.016497
4	21	0.016447
...
76	67	0.009870
77	71	0.009745
78	66	0.009650
79	64	0.008820
80	65	0.007599

81 rows × 2 columns

From the table above, we can see that part ID 23, 28, 32 have the top three largest total errors. For these five part ID, they may have highest probability of bad quality. To further analyse which sensor parameters most affect these three parts, I will calculate clustermap to visualize the influence of parameters and total_error.

In [41]:

```
# Read five bad quality parts
part23 = pd.read_csv('Scope0023.csv', encoding='unicode_escape')
part28 = pd.read_csv('Scope0028.csv', encoding='unicode_escape')
part32 = pd.read_csv('Scope0032.csv', encoding='unicode_escape')
```

For sensor machine parameters, there are 99 parameters in total. However, not all parameters' status is 'in use'. I assume that these 'not in use' and 'unknow parameters' are old fashioned sensors which are not used. In this case, I will only analyse quality based on sensor parameters which is 'in use' status.

In [42]:

```
# Check the parameter sheet
data = pd.ExcelFile('ForgedPartDataStructureSummaryv3.xlsx')
data_names = data.sheet_names
data_names
```

Out[42]:

```
['CMM Data Structure', 'Machine Parameters', 'Reordered Machine Parameters']
```


In [43]:

```
# Only get 'Machine Parameters' dataset
for par in data_names:
    if par == 'Machine Parameters':
        parameter = pd.read_excel('ForgedPartDataStructureSummaryv3.xlsx', sheet_name=parameter)
```

Out[43]:

	ID	Classification	Signal Name	Description	Nominal Value	Unit	Notes
0	1	In Use	Timer Tick [ms]	NaN	NaN	ms	NaN
1	2	NaN	Block-Nr	NaN	NaN	NaN	Interpretation of this variable is unknown
2	3	In Use	Power [kW]	Actual value of forging drive power	NaN	kW	NaN
3	4	In Use	Force [kN]	Hammer Force	NaN	kN	NaN
4	5	In Use	A_ges_vibr	NaN	NaN	NaN	Interpretation of this variable is unknown
...
94	95	Not changing	\$H1P_Y11 (U11S7)	NaN	NaN	NaN	Interpretation of this variable is unknown
95	96	Not in Use	\$U_GH_NOMEXT_2 (U26S1)	Nominal Heater 2	NaN	NaN	NaN
96	97	Not in Use	\$U_GH_HEATON_2 (U26S0)	Digital Signal to On/Off induction heater 2 (d...	NaN	NaN	NaN
97	98	Not in Use	\$U_GH_NOMEXT_1 (U25S1)	Nominal Heater 1	NaN	NaN	No variable data
98	99	In Use	\$U_GH_HEATON_1 (U25S0).1	Digital Signal to On/Off induction heater 1 (d...	NaN	Digital	Digital signal to indicate induction heating ...

99 rows × 7 columns

In [44]:

```
# Check columns of parameter datasets
parameter.columns
```

Out[44]:

```
Index(['ID', 'Classification', 'Signal Name', 'Description', 'Nominal Value',
      'Unit', 'Notes'],
      dtype='object')
```

In [45]:

```
# There are many sensor parameter status, I only choose parameter with 'In Use'
parameter['Classification'].unique()
```

Out[45]:

```
array(['In Use', nan, 'Auxiliary Process Measurement', 'Not in Use',
      'Not changing', 'Duplicate',
      'Valid (but does not contain information)'], dtype=object)
```

In [60]:

```
Find signal name which classification is 'in use'
signal_name = parameter[parameter['Classification'] == 'In Use'] # Select all 'In Use'
signal_name = signal_name[['Signal Name']] # only choose Signal Name column
signal_name
```

Out[60]:

	Signal Name
0	Timer Tick [ms]
2	Power [kW]
3	Force [kN]
4	A_ges_vibr
5	Schlagzahl [1/min]
6	EXZ_pos [deg]
9	A_ACTpos [mm]
11	DB_ACTpos [mm]
13	L_ACTpos [mm]
14	R_ACTpos [mm]

In [47]:

```
# get signal_name parameters
signal_name['Signal Name'].unique()
```

Out[47]:

```
array(['Timer Tick [ms]', 'Power [kW]', 'Force [kN]', 'A_ges_vibr',
      'Schlagzahl [1/min]', 'EXZ_pos [deg]', 'A_ACTpos [mm]',
      'DB_ACTpos [mm]', 'L_ACTpos [mm]', 'R_ACTpos [mm]',
      'SBA_ActPos [mm]', 'INDA_ACTpos [deg]', 'A_ACT_Force [kN]',
      'DB_ACT_Force [kN]', 'L_ACTspd [mm/min]', 'R_ACTspd [mm/min]',
      'SBA_NomPos [mm] [mm]', 'A_ACTspd [mm/min]', 'DB_ACTspd [mm/mi
n]',
      'L_NOMpos [mm]', 'R_NOMpos [mm]', 'SBA_OUT [%]', 'A_NOMpos [m
m]',
      'DB_NOMpos [mm]', 'L_OUT [%]', 'R_OUT [%]', 'Feedback SBA [%]',
      'A_OUT [%]', 'DB_OUT [%]', 'L_NOMspd [mm/min]',
      'R_NOMspd [mm/min]', 'Frc_Volt', 'A_NOMspd [mm/min]',
      'DB_NOMspd [mm/min]', 'Feedback L [%]', 'Feedback R [%]',
      'Speed Vn_1 [rpm]', 'NOMforceSPA [kN]', 'IP_ActSpd [mm/min]',
      'IP_ActPos [mm]', 'SPA_OUT [%]', 'Feedback A [%]',
      'Feedback DB [%]', 'IP_NomSpd [mm/min]', 'IP_NomPos',
      'Feedback_SPA [%]', 'ForgingBox_Temp', 'TMP_Ind_U1 [°C]',
      'TMP_Ind_F [°C]', 'IP_Out [%]', 'ACTforceSPA [kN]',
      '$U_GH_HEATON_1 (U25S0).1'], dtype=object)
```

Drop signal names which parameters are not 'In Use'

Part23

In [48]:

```
# Select datasets which only 'In Use' signal name parameters
part23 = part23.loc[:,['Timer Tick [ms]', 'Power [kW]', 'Force [kN]', 'A_ges_vibr',
'Schlagzahl [1/min]', 'EXZ_pos [deg]', 'A_ACTpos [mm]',
'DB_ACTpos [mm]', 'L_ACTpos [mm]', 'R_ACTpos [mm]',
'SBA_ActPos [mm]', 'INDA_ACTpos [deg]', 'A_ACT_Force [kN]',
'DB_ACT_Force [kN]', 'L_ACTspd [mm/min]', 'R_ACTspd [mm/min]',
'SBA_NomPos [mm]', 'A_ACTspd [mm/min]', 'DB_ACTspd [mm/min]',
'L_NOMpos [mm]', 'R_NOMpos [mm]', 'SBA_OUT [%]', 'A_NOMpos [mm]',
'DB_NOMpos [mm]', 'L_OUT [%]', 'R_OUT [%]', 'Feedback SBA [%]',
'A_OUT [%]', 'DB_OUT [%]', 'L_NOMspd [mm/min]',
'R_NOMspd [mm/min]', 'Frc_Volt', 'A_NOMspd [mm/min]',
'DB_NOMspd [mm/min]', 'Feedback L [%]', 'Feedback R [%]',
'Speed Vn_1 [rpm]', 'NOMforceSPA [kN]', 'IP_ActSpd [mm/min]',
'IP_ActPos [mm]', 'SPA_OUT [%]', 'Feedback A [%]',
'Feedback DB [%]', 'IP_NomSpd [mm/min]', 'IP_NomPos',
'Feedback_SPA [%]', 'ForgingBox_Temp', 'TMP_Ind_U1 [°C]',
'TMP_Ind_F [°C]', 'IP_Out [%]', 'ACTforceSPA [kN]',
'$U_GH_HEATON_1 (U25S0).1']]
```

part23

	[ms]	[kW]	[kN]	A_ges_vibr	[1/min]	[deg]	[mm]	[mm]	[mm]
0	2599023562	46.912811	0.0	5.971617	1207.17628	206.282891	1199.945	499.820	94.899990
1	2599023572	46.912830	0.0	5.971617	1207.83657	278.660820	1199.945	499.630	94.899992
2	2599023582	46.912993	0.0	5.211279	1207.21498	351.214531	1199.940	499.375	94.899988
3	2599023592	46.912733	0.0	5.313335	1206.20648	63.548516	1199.940	499.100	94.899978
4	2599023602	46.914011	0.0	5.687249	1206.18934	136.102227	1199.945	498.810	94.899967
...
17856	2599202122	46.765512	0.0	10.161914	1207.60470	270.311211	1199.925	500.000	94.900134
17857	2599202132	46.760248	0.0	10.161914	1206.48264	342.820977	1199.930	500.000	94.900141
17858	2599202142	46.755093	0.0	10.161914	1206.29344	55.111016	1199.930	500.000	94.900144
17859	2599202152	46.746974	0.0	10.161914	1206.63813	127.664727	1199.930	500.000	94.900148
17860	2599202162	46.736237	0.0	10.161914	1205.39208	199.954766	1199.930	500.000	94.900152

In [49]:

```
# As each value differ, I use normalization to rerange all values into [0,1] field
transfer = MinMaxScaler()
part23_new = transfer.fit_transform(part23) # Normalize part23 values
part23_new = pd.DataFrame(part23_new, columns = part23.columns) # Transfer to DataFrame
part23_error = cmm_part['total_error'].iloc[0] # Get the Part23's total error
part23_new['part23_error'] = part23_error # Create a new column for part 23's error
part23_new.head()
```

Out[49]:

	Timer Tick [ms]	Power [kW]	Force [kN]	A_ges_vibr	Schlagzahl [1/min]	EXZ_pos [deg]	A_ACTpos [mm]	DB_ACTpos [mm]	L_ACTpos [mm]
0	0.000000	0.013465	0.0	0.000402	0.670256	0.573068	0.999788	0.998589	0.999788
1	0.000056	0.013466	0.0	0.000402	0.722384	0.774142	0.999788	0.997615	0.999788
2	0.000112	0.013470	0.0	0.000351	0.673311	0.975705	0.999783	0.996306	0.999783
3	0.000168	0.013463	0.0	0.000358	0.593692	0.176535	0.999783	0.994896	0.999783
4	0.000224	0.013495	0.0	0.000383	0.592339	0.378098	0.999788	0.993408	0.999788

5 rows × 10 columns

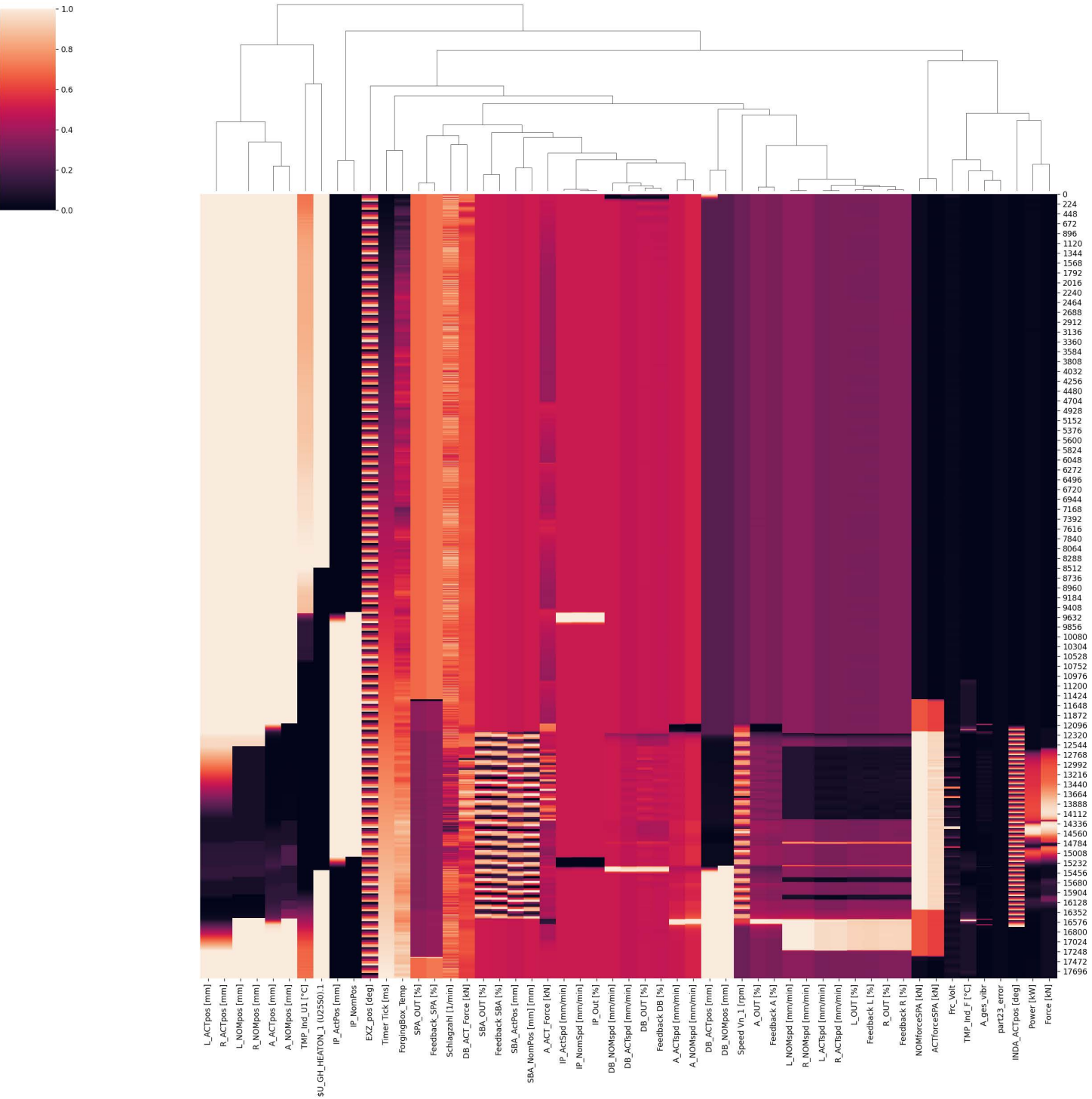
In [50]:

```
x,y= part23_new.iloc[:, 0:-1], part23_new.loc[:, 'part23_error']
```

In [51]:

```
sns.clustermap(data=part23_new, pivot_kws=None, method='average',
figsize=(5, 5), cbar_kws=None,
row_cluster=False, col_cluster=True)
```

The Clustermap of Part ID.23



In [52]: **Part28**

```
# Select datasets which only 'In Use' signal name parameters
part28 = part28.loc[:, ['Timer Tick [ms]', 'Power [kW]', 'Force [kN]', 'A_ges_vibr',
'Schlagzahl [1/min]', 'EXZ_pos [deg]', 'A_ACTpos [mm]',
'DB_ACTpos [mm]', 'L_ACTpos [mm]', 'R_ACTpos [mm]',
'SBA_ActPos [mm]', 'INDA_ACTpos [deg]', 'A_ACT_Force [kN]',
'DB_ACT_Force [kN]', 'L_ACTspd [mm/min]', 'R_ACTspd [mm/min]',
'SBA_NomPos [mm]', 'A_ACTspd [mm/min]', 'DB_ACTspd [mm/min]',
'L_NOMpos [mm]', 'R_NOMpos [mm]', 'SBA_OUT [%]', 'A_NOMpos [mm]',
'DB_NOMpos [mm]', 'L_OUT [%]', 'R_OUT [%]', 'Feedback SBA [%]',
'A_OUT [%]', 'DB_OUT [%]', 'L_NOMspd [mm/min]',
'R_NOMspd [mm/min]', 'Frc_Volt', 'A_NOMspd [mm/min]',
'DB_NOMspd [mm/min]', 'Feedback L [%]', 'Feedback R [%]',
'Speed Vn_1 [rpm]', 'NOMforceSPA [kN]', 'IP_ActSpd [mm/min]',
'IP_ActPos [mm]', 'SPA_OUT [%]', 'Feedback A [%]',
'Feedback DB [%]', 'IP_NomSpd [mm/min]', 'IP_NomPos',
'Feedback_SPA [%]', 'ForgingBox_Temp', 'TMP_Ind_U1 [°C]',
'TMP_Ind_F [°C]', 'IP_Out [%]', 'ACTforceSPA [kN]',
'$U_GH_HEATON_1 (U25S0).1']]

part28
```

Out[52]:

	Timer Tick [ms]	Power [kW]	Force [kN]	A_ges_vibr	Schlagzahl [1/min]	EXZ_pos [deg]	A_ACTpos [mm]	DB_ACTpos [mm]	L_ACTpos [mm]	F
0	2599919189	46.007757	0.0	5.753715	1205.25764	8.309258	1199.965	499.925	94.899937	9
1	2599919199	46.010688	0.0	5.753715	1206.04694	80.731133	1199.965	499.745	94.899938	9
2	2599919209	46.012026	0.0	5.753715	1205.00891	152.933281	1199.965	499.465	94.899938	9
3	2599919219	46.012928	0.0	5.112676	1204.83110	225.355156	1199.965	499.255	94.899939	9
4	2599919229	46.013953	0.0	0.536051	1204.60876	297.601250	1199.965	498.965	94.899943	9
...
17868	2600097869	45.919296	0.0	5.107856	1203.67931	31.072930	1199.965	500.000	94.899969	9
17869	2600097879	45.916606	0.0	5.107856	1204.42075	103.362969	1199.965	500.000	94.899973	9
17870	2600097889	45.910914	0.0	5.107856	1204.92432	175.521172	1199.965	500.000	94.899980	9

In [53]:

```
# As each value differ, I use normalization to rerange all values into [0,1] field
transfer = MinMaxScaler()
part28_new = transfer.fit_transform(part28) # Normalize part28 values
part28_new = pd.DataFrame(part28_new, columns = part23.columns) # Transfer columns
part28_error = cmm_part['total_error'].iloc[1] # Get the Part28's total error, which is 0.000000
part28_new['part28_error'] = part28_error # Create a new column for part 28's error
part28_new.head()
```

Out[53]:

	Timer Tick [ms]	Power [kW]	Force [kN]	A_ges_vibr	Schlagzahl [1/min]	EXZ_pos [deg]	A_ACTpos [mm]	DB_ACTpos [mm]	L_ACTpos [mm]
0	0.000000	0.013173	0.0	0.000408	0.725166	0.023074	0.999808	0.998718	0.999808
1	0.000056	0.013245	0.0	0.000408	0.789084	0.224271	0.999808	0.997794	0.999808
2	0.000112	0.013277	0.0	0.000408	0.705024	0.424857	0.999808	0.996358	0.999808
3	0.000168	0.013299	0.0	0.000363	0.690625	0.626053	0.999808	0.995281	0.999808
4	0.000224	0.013324	0.0	0.000038	0.672620	0.826761	0.999808	0.993793	0.999808

5 rows × 53 columns

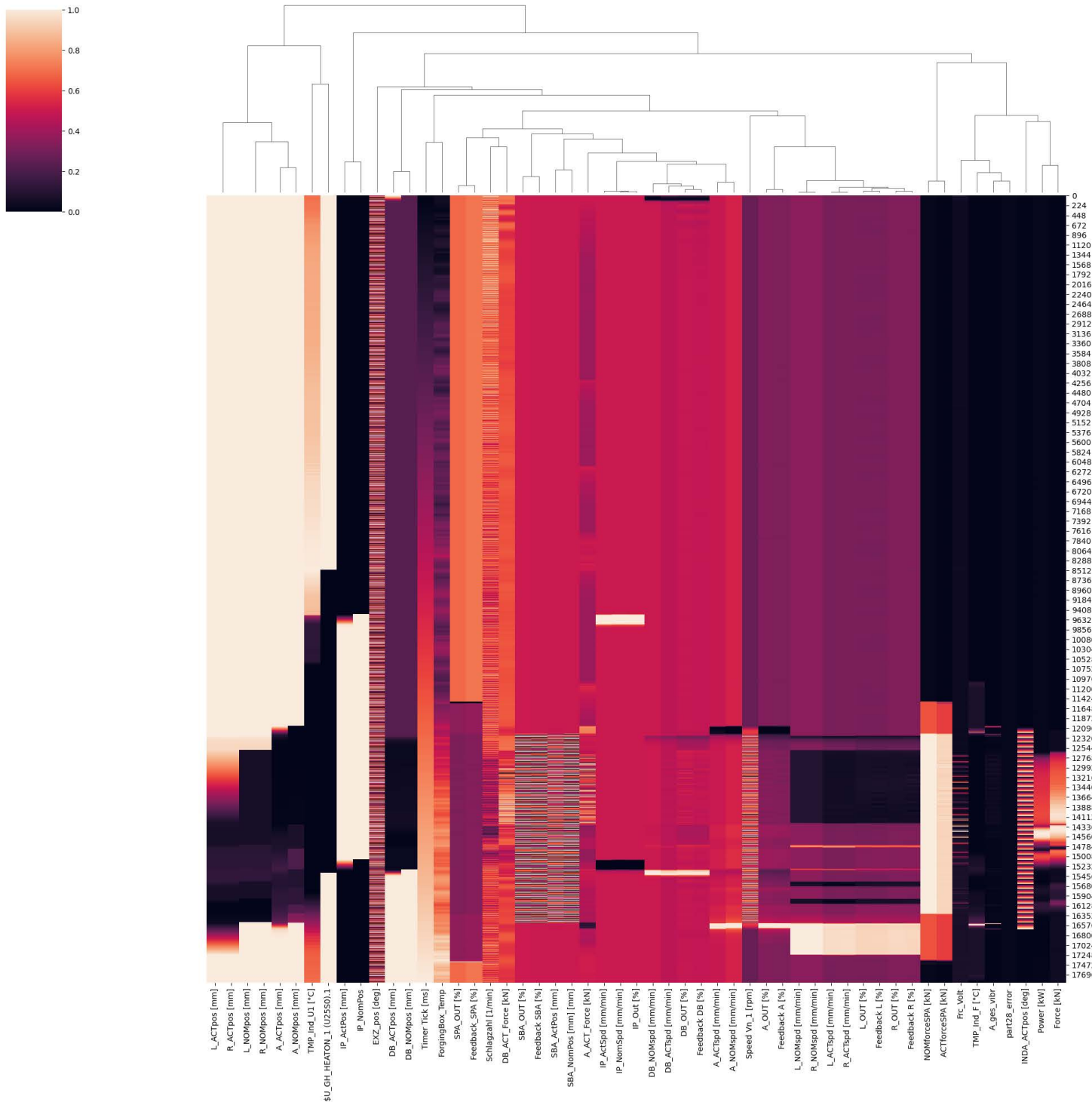
In [54]:

```
x,y= part28_new.iloc[:, 0:-1], part28_new.loc[:, 'part28_error']
```

In [55]:

```
sns.clustermmap(data=part28_new, pivot_kws=None, method='average',
                figsize=(5, 5), cbar_kws=None,
                row_cluster=False, col_cluster=True)
```


The Cluster of Part ID.28



In [56]:

```
# Select datasets which only 'In Use' signal name parameters
part32 = part32.loc[:,['Timer Tick [ms]', 'Power [kW]', 'Force [kN]', 'A_ges_vibr',
'Schlagzahl [1/min]', 'EXZ_pos [deg]', 'A_ACTpos [mm]',
'DB_ACTpos [mm]', 'L_ACTpos [mm]', 'R_ACTpos [mm]',
'SBA_ActPos [mm]', 'INDA_ACTpos [deg]', 'A_ACT_Force [kN]',
'DB_ACT_Force [kN]', 'L_ACTspd [mm/min]', 'R_ACTspd [mm/min]',
'SBA_NomPos [mm]', 'A_ACTspd [mm/min]', 'DB_ACTspd [mm/min]',
'L_NOMpos [mm]', 'R_NOMpos [mm]', 'SBA_OUT [%]', 'A_NOMpos [mm]',
'DB_NOMpos [mm]', 'L_OUT [%]', 'R_OUT [%]', 'Feedback SBA [%]',
'A_OUT [%]', 'DB_OUT [%]', 'L_NOMspd [mm/min]',
'R_NOMspd [mm/min]', 'Frc_Volt', 'A_NOMspd [mm/min]',
'DB_NOMspd [mm/min]', 'Feedback L [%]', 'Feedback R [%]',
'Speed Vn_1 [rpm]', 'NOMforceSPA [kN]', 'IP_ActSpd [mm/min]',
'IP_ActPos [mm]', 'SPA_OUT [%]', 'Feedback A [%]',
'Feedback DB [%]', 'IP_NomSpd [mm/min]', 'IP_NomPos',
'Feedback_SPA [%]', 'ForgingBox_Temp', 'TMP_Ind_U1 [°C]',
'TMP_Ind_F [°C]', 'IP_Out [%]', 'ACTforceSPA [kN]',
'$U_GH_HEATON_1 (U25S0).1']]

part32
```

Out[56]:

	Timer Tick [ms]	Power [kW]	Force [kN]	A_ges_vibr	Schlagzahl [1/min]	EXZ_pos [deg]	A_ACTpos [mm]	DB_ACTpos [mm]	L_ACTpo [mm]
0	2600637294	46.215480	0.000000	5.224182	1208.07322	183.035820	1199.975	499.980	94.90005
1	2600637304	46.216380	0.000000	5.429102	1207.52203	255.589531	1199.980	499.895	94.90005
2	2600637314	46.219487	0.000000	5.429102	1206.78169	327.967461	1199.980	499.725	94.90004
3	2600637324	46.220170	0.000000	5.429102	1205.83432	40.477227	1199.980	499.440	94.90004
4	2600637334	46.222089	0.000000	5.429102	1206.89631	113.030937	1199.980	499.220	94.90004
...
17869	2600815984	46.420413	34.599609	0.068260	1212.41813	137.772148	1199.955	500.000	94.89996
17870	2600815994	46.420047	34.599609	0.076315	1211.92815	210.545586	1199.955	500.000	94.89996
17871	2600816004	46.421076	34.599609	0.083133	1211.00230	283.055352	1199.955	500.000	94.89996

In [57]:

```
# As each value differ, I use normalization to rerange all values into [0,1] field
transfer = MinMaxScaler()
part32_new = transfer.fit_transform(part32) # Normalize part32 values
part32_new = pd.DataFrame(part32_new, columns = part32.columns) # Transfer
part32_error = cmm_part['total_error'].iloc[2] # Get the Part32's total error, which is 0.000362
part32_new['part32_error'] = part32_error # Create a new column for part 32's error
part32_new.head()
```

Out[57]:

	Timer Tick [ms]	Power [kW]	Force [kN]	A_ges_vibr	Schlagzahl [1/min]	EXZ_pos [deg]	A_ACTpos [mm]	DB_ACTpos [mm]	L_ACTpos [mm]
0	0.000000	0.004504	0.0	0.000362	0.364892	0.508427	0.999818	0.999026	0.999026
1	0.000056	0.004526	0.0	0.000376	0.315285	0.710064	0.999823	0.998590	0.999026
2	0.000112	0.004601	0.0	0.000376	0.248655	0.911212	0.999823	0.997718	0.999026
3	0.000168	0.004618	0.0	0.000376	0.163393	0.112237	0.999823	0.996257	0.999026
4	0.000224	0.004665	0.0	0.000376	0.258971	0.313874	0.999823	0.995128	0.999026

5 rows × 10 columns

In [58]:

```
x,y= part32_new.iloc[:, 0:-1], part32_new.loc[:, 'part32_error']
```

In [59]:

```
sns.clustermap(data=part32_new, pivot_kws=None,
               figsize=(5, 5), cbar_kws=None,
               row_cluster=False, col_cluster=True)
```

The Cluster of Part ID.32

