

# Arcing\_Issue\_Case\_Random\_Forest-20221109

November 9, 2022

## 1 Arcing Issue

```
[117]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

### 1.1 import data

```
[118]: dataset=r'C:\Users\pj007\exercises\JMP_Case\dataset\Arcing_Issue_195.csv'
data=pd.read_csv(dataset)
data
```

```
[118]:
```

	Glass ID	Arcing Classification	1PC_ITEMP_AVG	1PC_ITEMP_MAX	\
0	G1	Y	339.7	340.0	
1	G2	N	341.4	341.6	
2	G4	N	340.8	341.0	
3	G5	Y	339.4	339.6	
4	G6	N	340.9	341.0	
5	G10	N	339.0	339.0	
6	G11	N	341.9	342.1	
7	G14	Y	339.5	339.7	
8	G15	Y	340.7	340.8	
9	G16	Y	339.1	339.5	
10	G19	Y	340.5	340.7	
11	G20	N	341.1	341.3	
12	G21	Y	338.4	338.6	
13	G23	Y	339.4	339.7	
14	G24	N	339.2	339.5	
15	G25	N	340.7	341.0	
16	G26	Y	339.2	339.3	
17	G27	N	340.6	340.6	
18	G28	Y	338.6	338.8	
19	G30	Y	339.4	339.6	
20	G36	Y	339.8	340.2	
21	G37	Y	340.7	340.8	
22	G38	Y	339.6	339.7	
23	G39	N	341.2	341.3	

24	G3	Y	339.1	339.5
25	G7	Y	339.3	339.7
26	G13	Y	341.8	342.0
27	G17	N	340.9	341.0
28	G18	Y	339.1	339.3
29	G22	N	340.4	340.6
30	G33	N	339.7	339.8
31	G34	N	340.8	341.0
32	G8	N	341.0	341.0
33	G9	Y	339.0	339.2
34	G12	Y	339.9	340.2
35	G29	Y	341.6	342.0
36	G31	Y	338.6	338.8
37	G32	N	341.7	342.0
38	G35	Y	340.4	340.7
39	G40	Y	339.0	339.2

	1PC_ITEMP_MIN	1PC_OTEMP2_AVG	1PC_OTEMP2_MAX	1PC_OTEMP2_MIN	\
0	339.5	360.1	360.2	359.9	
1	341.3	360.2	360.2	360.2	
2	340.7	360.2	360.3	360.2	
3	339.1	360.2	360.3	360.0	
4	340.7	359.8	360.2	359.4	
5	339.0	359.8	360.2	359.5	
6	341.8	359.9	360.4	359.7	
7	339.2	359.6	359.8	359.5	
8	340.7	359.6	359.9	359.3	
9	338.8	360.0	360.2	359.9	
10	340.5	359.8	360.0	359.4	
11	341.0	359.9	360.0	359.8	
12	338.2	359.6	359.9	359.3	
13	339.2	359.8	360.0	359.5	
14	339.0	360.1	360.3	359.9	
15	340.7	360.1	360.3	359.8	
16	339.0	360.2	360.3	360.2	
17	340.5	360.2	360.3	359.9	
18	338.3	360.2	360.2	360.2	
19	339.2	359.6	359.7	359.5	
20	339.6	359.5	359.7	359.4	
21	340.5	360.0	360.3	359.7	
22	339.3	359.6	359.7	359.5	
23	341.1	360.0	360.0	360.0	
24	338.8	360.1	360.2	359.9	
25	339.2	359.7	359.9	359.3	
26	341.6	359.8	359.9	359.7	
27	340.8	359.5	360.0	359.3	
28	338.8	359.7	360.0	359.4	

29	340.3	359.6	360.0	359.4
30	339.6	359.7	359.8	359.7
31	340.6	359.7	360.0	359.5
32	341.0	359.7	359.9	359.5
33	338.7	359.6	360.0	359.4
34	339.7	359.7	360.2	359.4
35	341.5	359.5	359.8	359.4
36	338.3	359.9	360.3	359.5
37	341.6	359.8	360.0	359.7
38	340.2	360.2	360.3	360.0
39	338.7	360.2	360.4	359.9

	1PC_OTEMP_AVG	1PC_OTEMP_MAX	...	6RF_VDC_MAX	6RF_VDC_MIN	6RF_VPP_AVG	\
0	359.6	360.0	...	20.9	20.8	1767.2	
1	359.6	359.9	...	20.8	20.6	1768.4	
2	359.6	360.0	...	20.8	20.8	1756.6	
3	359.6	360.0	...	20.8	20.8	1750.9	
4	359.6	359.9	...	20.8	20.7	1759.1	
5	359.9	359.9	...	21.0	21.0	1769.7	
6	359.7	359.8	...	21.0	20.9	1766.5	
7	360.0	360.4	...	20.9	20.8	1759.1	
8	359.7	360.0	...	20.9	20.9	1750.3	
9	360.2	360.2	...	20.9	20.8	1747.2	
10	359.6	359.8	...	21.0	20.9	1739.1	
11	360.2	360.3	...	20.7	20.7	1749.1	
12	359.4	359.5	...	20.9	20.8	1763.4	
13	359.6	359.9	...	20.8	20.8	1754.1	
14	360.0	360.3	...	20.9	20.9	1754.7	
15	359.8	360.0	...	20.8	20.8	1751.6	
16	359.6	359.8	...	20.8	20.8	1746.0	
17	359.5	359.9	...	21.0	20.8	1742.8	
18	359.8	360.2	...	20.8	20.8	1764.7	
19	359.9	360.0	...	20.8	20.8	1756.6	
20	360.3	360.4	...	20.9	20.9	1775.9	
21	359.6	359.9	...	21.0	20.9	1752.8	
22	360.1	360.2	...	20.9	20.8	1760.3	
23	359.4	359.4	...	20.9	20.9	1760.3	
24	359.5	359.7	...	20.9	20.7	1762.2	
25	359.5	359.7	...	21.0	20.8	1755.9	
26	360.2	360.3	...	20.9	20.7	1761.6	
27	359.5	359.7	...	20.9	20.8	1746.6	
28	359.6	359.7	...	21.0	20.9	1745.3	
29	359.6	360.0	...	21.0	20.9	1747.2	
30	360.2	360.4	...	20.9	20.8	1755.9	
31	359.7	360.2	...	21.1	21.0	1754.1	
32	359.5	359.7	...	20.9	20.9	1750.3	
33	359.5	359.8	...	20.8	20.8	1749.1	

34	359.7	359.9	...	20.8	20.8	1766.5
35	360.1	360.3	...	20.8	20.8	1761.6
36	359.7	360.2	...	20.9	20.9	1771.5
37	360.2	360.3	...	20.9	20.8	1767.8
38	359.8	360.2	...	21.1	21.1	1770.9
39	359.6	359.9	...	20.9	20.9	1755.9

	6RF_VPP_MAX	6RF_VPP_MIN	6TH_VALVE_AVG	6TH_VALVE_MAX	6TH_VALVE_MIN	\
0	1778.4	1759.7	14.4	16.2	12.2	
1	1782.1	1755.9	15.1	16.8	13.2	
2	1765.3	1748.4	12.8	15.5	9.6	
3	1761.6	1742.8	14.2	15.8	12.0	
4	1772.8	1746.6	15.1	16.3	13.8	
5	1778.4	1763.4	15.1	20.2	9.7	
6	1774.7	1759.7	13.8	16.5	9.6	
7	1770.9	1750.3	15.4	16.9	13.9	
8	1759.7	1742.8	14.8	16.5	13.0	
9	1757.8	1737.2	13.8	15.7	11.9	
10	1748.4	1731.6	14.5	16.1	12.6	
11	1761.6	1737.2	15.4	16.7	14.3	
12	1770.9	1757.8	15.0	17.2	12.2	
13	1767.2	1742.8	15.1	15.9	14.7	
14	1767.2	1744.7	13.3	15.9	10.0	
15	1763.4	1742.8	15.8	17.9	13.8	
16	1755.9	1737.2	14.3	16.0	12.6	
17	1754.1	1731.6	14.2	15.3	12.1	
18	1776.5	1755.9	14.4	16.1	11.8	
19	1767.2	1746.6	15.2	17.4	13.3	
20	1785.9	1767.2	14.8	16.1	13.2	
21	1763.4	1742.8	13.2	15.7	9.8	
22	1770.9	1750.3	15.0	16.4	13.4	
23	1770.9	1752.2	15.5	16.7	14.0	
24	1772.8	1754.1	14.8	16.3	12.9	
25	1765.3	1746.6	15.0	16.5	14.0	
26	1772.8	1754.1	15.1	16.4	13.6	
27	1757.8	1737.2	15.1	16.5	13.6	
28	1757.8	1735.3	14.5	15.9	12.0	
29	1755.9	1741.0	13.1	15.1	9.5	
30	1765.3	1746.6	13.4	16.2	9.2	
31	1765.3	1744.7	14.7	16.2	12.9	
32	1763.4	1739.1	15.2	16.5	13.3	
33	1761.6	1739.1	15.4	17.1	13.2	
34	1778.4	1755.9	15.2	16.0	14.4	
35	1772.8	1752.2	14.2	16.3	12.0	
36	1782.1	1763.4	14.3	17.3	10.4	
37	1778.4	1759.7	14.5	16.4	12.1	
38	1782.1	1761.6	14.8	16.2	13.0	

39	1765.3	1748.4	14.6	16.3	12.7
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	6WFS_PM
0	1.0
1	2.0
2	4.0
3	5.0
4	4.0
5	1.0
6	2.0
7	3.0
8	4.0
9	5.0
10	8.0
11	8.0
12	1.0
13	5.0
14	5.0
15	6.0
16	7.0
17	8.0
18	1.0
19	3.0
20	2.0
21	5.0
22	6.0
23	7.0
24	3.0
25	5.0
26	2.0
27	6.0
28	7.0
29	4.0
30	3.0
31	4.0
32	6.0
33	7.0
34	3.0
35	2.0
36	1.0
37	2.0
38	1.0
39	8.0

[40 rows x 198 columns]

### 1.1.1 train\_set

```
[119]: train_data=data.iloc[0:32,2:-1]
train_data
```

```
[119]:
```

	1PC_ITEMP_AVG	1PC_ITEMP_MAX	1PC_ITEMP_MIN	1PC_OTEMP2_AVG	\
0	339.7	340.0	339.5	360.1	
1	341.4	341.6	341.3	360.2	
2	340.8	341.0	340.7	360.2	
3	339.4	339.6	339.1	360.2	
4	340.9	341.0	340.7	359.8	
5	339.0	339.0	339.0	359.8	
6	341.9	342.1	341.8	359.9	
7	339.5	339.7	339.2	359.6	
8	340.7	340.8	340.7	359.6	
9	339.1	339.5	338.8	360.0	
10	340.5	340.7	340.5	359.8	
11	341.1	341.3	341.0	359.9	
12	338.4	338.6	338.2	359.6	
13	339.4	339.7	339.2	359.8	
14	339.2	339.5	339.0	360.1	
15	340.7	341.0	340.7	360.1	
16	339.2	339.3	339.0	360.2	
17	340.6	340.6	340.5	360.2	
18	338.6	338.8	338.3	360.2	
19	339.4	339.6	339.2	359.6	
20	339.8	340.2	339.6	359.5	
21	340.7	340.8	340.5	360.0	
22	339.6	339.7	339.3	359.6	
23	341.2	341.3	341.1	360.0	
24	339.1	339.5	338.8	360.1	
25	339.3	339.7	339.2	359.7	
26	341.8	342.0	341.6	359.8	
27	340.9	341.0	340.8	359.5	
28	339.1	339.3	338.8	359.7	
29	340.4	340.6	340.3	359.6	
30	339.7	339.8	339.6	359.7	
31	340.8	341.0	340.6	359.7	

	1PC_OTEMP2_MAX	1PC_OTEMP2_MIN	1PC_OTEMP_AVG	1PC_OTEMP_MAX	\
0	360.2	359.9	359.6	360.0	
1	360.2	360.2	359.6	359.9	
2	360.3	360.2	359.6	360.0	
3	360.3	360.0	359.6	360.0	
4	360.2	359.4	359.6	359.9	
5	360.2	359.5	359.9	359.9	
6	360.4	359.7	359.7	359.8	

7	359.8	359.5	360.0	360.4
8	359.9	359.3	359.7	360.0
9	360.2	359.9	360.2	360.2
10	360.0	359.4	359.6	359.8
11	360.0	359.8	360.2	360.3
12	359.9	359.3	359.4	359.5
13	360.0	359.5	359.6	359.9
14	360.3	359.9	360.0	360.3
15	360.3	359.8	359.8	360.0
16	360.3	360.2	359.6	359.8
17	360.3	359.9	359.5	359.9
18	360.2	360.2	359.8	360.2
19	359.7	359.5	359.9	360.0
20	359.7	359.4	360.3	360.4
21	360.3	359.7	359.6	359.9
22	359.7	359.5	360.1	360.2
23	360.0	360.0	359.4	359.4
24	360.2	359.9	359.5	359.7
25	359.9	359.3	359.5	359.7
26	359.9	359.7	360.2	360.3
27	360.0	359.3	359.5	359.7
28	360.0	359.4	359.6	359.7
29	360.0	359.4	359.6	360.0
30	359.8	359.7	360.2	360.4
31	360.0	359.5	359.7	360.2

	1PC_OTEMP_MIN	1PRESSURE_AVG	...	6RF_VDC_AVG	6RF_VDC_MAX	6RF_VDC_MIN	\
0	359.3	7.8	...	20.9	20.9	20.8	
1	359.5	8.7	...	20.7	20.8	20.6	
2	359.4	7.8	...	20.8	20.8	20.8	
3	359.4	7.8	...	20.8	20.8	20.8	
4	359.3	8.3	...	20.8	20.8	20.7	
5	359.8	8.8	...	21.0	21.0	21.0	
6	359.5	8.5	...	21.0	21.0	20.9	
7	359.7	9.3	...	20.8	20.9	20.8	
8	359.3	8.7	...	20.9	20.9	20.9	
9	360.0	8.5	...	20.9	20.9	20.8	
10	359.3	8.5	...	21.0	21.0	20.9	
11	359.9	8.2	...	20.7	20.7	20.7	
12	359.3	8.6	...	20.8	20.9	20.8	
13	359.4	8.6	...	20.8	20.8	20.8	
14	359.7	8.5	...	20.9	20.9	20.9	
15	359.4	8.3	...	20.8	20.8	20.8	
16	359.4	8.5	...	20.8	20.8	20.8	
17	359.4	8.3	...	20.9	21.0	20.8	
18	359.4	9.0	...	20.8	20.8	20.8	
19	359.7	9.0	...	20.8	20.8	20.8	

20	359.9	5.8	...	20.9	20.9	20.9
21	359.4	5.2	...	20.9	21.0	20.9
22	359.9	5.0	...	20.9	20.9	20.8
23	359.3	4.8	...	20.9	20.9	20.9
24	359.3	8.1	...	20.8	20.9	20.7
25	359.4	8.5	...	20.8	21.0	20.8
26	360.2	8.5	...	20.9	20.9	20.7
27	359.4	8.7	...	20.9	20.9	20.8
28	359.4	8.3	...	21.0	21.0	20.9
29	359.4	8.7	...	20.9	21.0	20.9
30	359.9	8.7	...	20.9	20.9	20.8
31	359.5	8.4	...	21.0	21.1	21.0

	6RF_VPP_AVG	6RF_VPP_MAX	6RF_VPP_MIN	6TH_VALVE_AVG	6TH_VALVE_MAX	\
0	1767.2	1778.4	1759.7	14.4	16.2	
1	1768.4	1782.1	1755.9	15.1	16.8	
2	1756.6	1765.3	1748.4	12.8	15.5	
3	1750.9	1761.6	1742.8	14.2	15.8	
4	1759.1	1772.8	1746.6	15.1	16.3	
5	1769.7	1778.4	1763.4	15.1	20.2	
6	1766.5	1774.7	1759.7	13.8	16.5	
7	1759.1	1770.9	1750.3	15.4	16.9	
8	1750.3	1759.7	1742.8	14.8	16.5	
9	1747.2	1757.8	1737.2	13.8	15.7	
10	1739.1	1748.4	1731.6	14.5	16.1	
11	1749.1	1761.6	1737.2	15.4	16.7	
12	1763.4	1770.9	1757.8	15.0	17.2	
13	1754.1	1767.2	1742.8	15.1	15.9	
14	1754.7	1767.2	1744.7	13.3	15.9	
15	1751.6	1763.4	1742.8	15.8	17.9	
16	1746.0	1755.9	1737.2	14.3	16.0	
17	1742.8	1754.1	1731.6	14.2	15.3	
18	1764.7	1776.5	1755.9	14.4	16.1	
19	1756.6	1767.2	1746.6	15.2	17.4	
20	1775.9	1785.9	1767.2	14.8	16.1	
21	1752.8	1763.4	1742.8	13.2	15.7	
22	1760.3	1770.9	1750.3	15.0	16.4	
23	1760.3	1770.9	1752.2	15.5	16.7	
24	1762.2	1772.8	1754.1	14.8	16.3	
25	1755.9	1765.3	1746.6	15.0	16.5	
26	1761.6	1772.8	1754.1	15.1	16.4	
27	1746.6	1757.8	1737.2	15.1	16.5	
28	1745.3	1757.8	1735.3	14.5	15.9	
29	1747.2	1755.9	1741.0	13.1	15.1	
30	1755.9	1765.3	1746.6	13.4	16.2	
31	1754.1	1765.3	1744.7	14.7	16.2	



	6TH_VALVE_MIN	6WFS_PM
0	12.2	1.0
1	13.2	2.0
2	9.6	4.0
3	12.0	5.0
4	13.8	4.0
5	9.7	1.0
6	9.6	2.0
7	13.9	3.0
8	13.0	4.0
9	11.9	5.0
10	12.6	8.0
11	14.3	8.0
12	12.2	1.0
13	14.7	5.0
14	10.0	5.0
15	13.8	6.0
16	12.6	7.0
17	12.1	8.0
18	11.8	1.0
19	13.3	3.0
20	13.2	2.0
21	9.8	5.0
22	13.4	6.0
23	14.0	7.0
24	12.9	3.0
25	14.0	5.0
26	13.6	2.0
27	13.6	6.0
28	12.0	7.0
29	9.5	4.0
30	9.2	3.0
31	12.9	4.0

[32 rows x 195 columns]

```
[120]: train_labels=data.iloc[0:32,1]
train_labels
```

```
[120]: 0    Y
1    N
2    N
3    Y
4    N
5    N
6    N
7    Y
```

```

8      Y
9      Y
10     Y
11     N
12     Y
13     Y
14     N
15     N
16     Y
17     N
18     Y
19     Y
20     Y
21     Y
22     Y
23     N
24     Y
25     Y
26     Y
27     N
28     Y
29     N
30     N
31     N
Name: Arcing Classification, dtype: object

```

### 1.1.2 test\_set

```
[121]: test_data=data.iloc[32:,-1]
test_data
```

```
[121]:
```

	1PC_ITEMP_AVG	1PC_ITEMP_MAX	1PC_ITEMP_MIN	1PC_OTEMP2_AVG	\
32	341.0	341.0	341.0	359.7	
33	339.0	339.2	338.7	359.6	
34	339.9	340.2	339.7	359.7	
35	341.6	342.0	341.5	359.5	
36	338.6	338.8	338.3	359.9	
37	341.7	342.0	341.6	359.8	
38	340.4	340.7	340.2	360.2	
39	339.0	339.2	338.7	360.2	

	1PC_OTEMP2_MAX	1PC_OTEMP2_MIN	1PC_OTEMP_AVG	1PC_OTEMP_MAX	\
32	359.9	359.5	359.5	359.7	
33	360.0	359.4	359.5	359.8	
34	360.2	359.4	359.7	359.9	
35	359.8	359.4	360.1	360.3	
36	360.3	359.5	359.7	360.2	

37	360.0	359.7	360.2	360.3
38	360.3	360.0	359.8	360.2
39	360.4	359.9	359.6	359.9

	1PC_OTEMP_MIN	1PRESSURE_AVG	...	6RF_VDC_AVG	6RF_VDC_MAX	6RF_VDC_MIN	\
32	359.4	8.1	...	20.9	20.9	20.9	
33	359.4	7.9	...	20.8	20.8	20.8	
34	359.7	8.6	...	20.8	20.8	20.8	
35	359.8	8.5	...	20.8	20.8	20.8	
36	359.4	8.5	...	20.9	20.9	20.9	
37	360.0	8.1	...	20.9	20.9	20.8	
38	359.5	5.8	...	21.1	21.1	21.1	
39	359.5	5.5	...	20.9	20.9	20.9	

	6RF_VPP_AVG	6RF_VPP_MAX	6RF_VPP_MIN	6TH_VALVE_AVG	6TH_VALVE_MAX	\
32	1750.3	1763.4	1739.1	15.2	16.5	
33	1749.1	1761.6	1739.1	15.4	17.1	
34	1766.5	1778.4	1755.9	15.2	16.0	
35	1761.6	1772.8	1752.2	14.2	16.3	
36	1771.5	1782.1	1763.4	14.3	17.3	
37	1767.8	1778.4	1759.7	14.5	16.4	
38	1770.9	1782.1	1761.6	14.8	16.2	
39	1755.9	1765.3	1748.4	14.6	16.3	

	6TH_VALVE_MIN	6WFS_PM
32	13.3	6.0
33	13.2	7.0
34	14.4	3.0
35	12.0	2.0
36	10.4	1.0
37	12.1	2.0
38	13.0	1.0
39	12.7	8.0

[8 rows x 195 columns]

```
[122]: test_labels=data.iloc[32:,1]
       test_labels
```

```
[122]: 32    N
       33    Y
       34    Y
       35    Y
       36    Y
       37    N
       38    Y
       39    Y
```

Name: Arcing Classification, dtype: object

## 1.2 export standard dataset

```
[123]: from sklearn import datasets, preprocessing
        #from sklearn.model_selection import train_test_split
        from sklearn.metrics import accuracy_score
        scaler=preprocessing.StandardScaler().fit(train_data)
        train_data=scaler.transform(train_data)
        test_data=scaler.transform(test_data)

[124]: Arcing_Issue_standard1=data.iloc[:,0:2]
        #Arcing_Issue_standard1

[125]: train_data_df=pd.DataFrame(train_data)
        #train_data_df

[126]: test_data_df=pd.DataFrame(test_data)
        #test_data_df

[127]: Arcing_Issue_standard2=pd.concat([train_data_df,test_data_df]).
        ↪reset_index(drop=True)
        #Arcing_Issue_standard2

[128]: Arcing_Issue_standard3=data.iloc[:, -1:]
        #Arcing_Issue_standard3

[129]: cols=list(data.columns)
        #cols

[130]: Arcing_Issue_standard=pd.
        ↪concat([Arcing_Issue_standard1,Arcing_Issue_standard2,Arcing_Issue_standard3],axis=1)
        #Arcing_Issue_standard

[131]: Arcing_Issue_standard.columns=cols
        Arcing_Issue_standard

[131]:
```

	Glass ID	Arcing Classification	1PC_ITEMP_AVG	1PC_ITEMP_MAX	\
0	G1	Y	-0.384757	-0.284609	
1	G2	N	1.435313	1.471051	
2	G4	N	0.792935	0.812679	
3	G5	Y	-0.705946	-0.723524	
4	G6	N	0.899998	0.812679	
5	G10	N	-1.134198	-1.381897	
6	G11	N	1.970627	2.019695	
7	G14	Y	-0.598883	-0.613795	

8	G15	Y	0.685872	0.593221
9	G16	Y	-1.027135	-0.833253
10	G19	Y	0.471746	0.483492
11	G20	N	1.114124	1.141865
12	G21	Y	-1.776576	-1.820812
13	G23	Y	-0.705946	-0.613795
14	G24	N	-0.920072	-0.833253
15	G25	N	0.685872	0.812679
16	G26	Y	-0.920072	-1.052710
17	G27	N	0.578809	0.373764
18	G28	Y	-1.562450	-1.601354
19	G30	Y	-0.705946	-0.723524
20	G36	Y	-0.277694	-0.065151
21	G37	Y	0.685872	0.593221
22	G38	Y	-0.491820	-0.613795
23	G39	N	1.221187	1.141865
24	G3	Y	-1.027135	-0.833253
25	G7	Y	-0.813009	-0.613795
26	G13	Y	1.863564	1.909966
27	G17	N	0.899998	0.812679
28	G18	Y	-1.027135	-1.052710
29	G22	N	0.364683	0.373764
30	G33	N	-0.384757	-0.504067
31	G34	N	0.792935	0.812679
32	G8	N	1.007061	0.812679
33	G9	Y	-1.134198	-1.162439
34	G12	Y	-0.170632	-0.065151
35	G29	Y	1.649438	1.909966
36	G31	Y	-1.562450	-1.601354
37	G32	N	1.756501	1.909966
38	G35	Y	0.364683	0.483492
39	G40	Y	-1.134198	-1.162439

	1PC_ITEMP_MIN	1PC_OTEMP2_AVG	1PC_OTEMP2_MAX	1PC_OTEMP2_MIN \
0	-0.399929	0.983611	0.649002	0.739290
1	1.428320	1.408957	0.649002	1.767868
2	0.818903	1.408957	1.143480	1.767868
3	-0.806207	1.408957	1.143480	1.082150
4	0.818903	-0.292425	0.649002	-0.975006
5	-0.907776	-0.292425	0.649002	-0.632147
6	1.936166	0.132920	1.637957	0.053572
7	-0.704638	-1.143116	-1.328909	-0.632147
8	0.818903	-1.143116	-0.834431	-1.317866
9	-1.110915	0.558266	0.649002	0.739290
10	0.615764	-0.292425	-0.339953	-0.975006
11	1.123611	0.132920	-0.339953	0.396431
12	-1.720332	-1.143116	-0.834431	-1.317866

13	-0.704638	-0.292425	-0.339953	-0.632147
14	-0.907776	0.983611	1.143480	0.739290
15	0.818903	0.983611	1.143480	0.396431
16	-0.907776	1.408957	1.143480	1.767868
17	0.615764	1.408957	1.143480	0.739290
18	-1.618762	1.408957	0.649002	1.767868
19	-0.704638	-1.143116	-1.823386	-0.632147
20	-0.298360	-1.568461	-1.823386	-0.975006
21	0.615764	0.558266	1.143480	0.053572
22	-0.603068	-1.143116	-1.823386	-0.632147
23	1.225181	0.558266	-0.339953	1.082150
24	-1.110915	0.983611	0.649002	0.739290
25	-0.704638	-0.717770	-0.834431	-1.317866
26	1.733028	-0.292425	-0.834431	0.053572
27	0.920473	-1.568461	-0.339953	-1.317866
28	-1.110915	-0.717770	-0.339953	-0.975006
29	0.412626	-1.143116	-0.339953	-0.975006
30	-0.298360	-0.717770	-1.328909	0.053572
31	0.717334	-0.717770	-0.339953	-0.632147
32	1.123611	-0.717770	-0.834431	-0.632147
33	-1.212485	-1.143116	-0.339953	-0.975006
34	-0.196791	-0.717770	0.649002	-0.975006
35	1.631458	-1.568461	-1.328909	-0.975006
36	-1.618762	0.132920	1.143480	-0.632147
37	1.733028	-0.292425	-0.339953	0.053572
38	0.311056	1.408957	1.143480	1.082150
39	-1.212485	1.408957	1.637957	0.739290

	1PC_OTEMP_AVG	1PC_OTEMP_MAX	...	6RF_VDC_MAX	6RF_VDC_MIN	6RF_VPP_AVG	\
0	-0.589421	0.074102	...	0.035377	-0.255434	1.311505	
1	-0.589421	-0.321110	...	-1.096701	-2.590827	1.453289	
2	-0.589421	0.074102	...	-1.096701	-0.255434	0.059077	
3	-0.589421	0.074102	...	-1.096701	-0.255434	-0.614399	
4	-0.589421	-0.321110	...	-1.096701	-1.423130	0.354461	
5	0.565363	-0.321110	...	1.167456	2.079960	1.606889	
6	-0.204493	-0.716322	...	1.167456	0.912263	1.228797	
7	0.950291	1.654952	...	0.035377	-0.255434	0.354461	
8	-0.204493	0.074102	...	0.035377	0.912263	-0.685291	
9	1.720147	0.864527	...	0.035377	-0.255434	-1.051567	
10	-0.589421	-0.716322	...	1.167456	0.912263	-2.008611	
11	1.720147	1.259739	...	-2.228780	-1.423130	-0.827075	
12	-1.359277	-1.901959	...	0.035377	-0.255434	0.862521	
13	-0.589421	-0.321110	...	-1.096701	-0.255434	-0.236307	
14	0.950291	1.259739	...	0.035377	0.912263	-0.165415	
15	0.180435	0.074102	...	-1.096701	-0.255434	-0.531691	
16	-0.589421	-0.716322	...	-1.096701	-0.255434	-1.193351	
17	-0.974349	-0.321110	...	1.167456	-0.255434	-1.571442	

18	0.180435	0.864527	...	-1.096701	-0.255434	1.016121
19	0.565363	0.074102	...	-1.096701	-0.255434	0.059077
20	2.105075	1.654952	...	0.035377	0.912263	2.339441
21	-0.589421	-0.321110	...	1.167456	0.912263	-0.389907
22	1.335219	0.864527	...	0.035377	-0.255434	0.496245
23	-1.359277	-2.297172	...	0.035377	0.912263	0.496245
24	-0.974349	-1.111535	...	0.035377	-1.423130	0.720737
25	-0.974349	-1.111535	...	1.167456	-0.255434	-0.023631
26	1.720147	1.259739	...	0.035377	-1.423130	0.649845
27	-0.974349	-1.111535	...	0.035377	-0.255434	-1.122459
28	-0.589421	-1.111535	...	1.167456	0.912263	-1.276059
29	-0.589421	0.074102	...	1.167456	0.912263	-1.051567
30	1.720147	1.654952	...	0.035377	-0.255434	-0.023631
31	-0.204493	0.864527	...	2.299535	2.079960	-0.236307
32	-0.974349	-1.111535	...	0.035377	0.912263	-0.685291
33	-0.974349	-0.716322	...	-1.096701	-0.255434	-0.827075
34	-0.204493	-0.321110	...	-1.096701	-0.255434	1.228797
35	1.335219	1.259739	...	-1.096701	-0.255434	0.649845
36	-0.204493	0.864527	...	0.035377	0.912263	1.819565
37	1.720147	1.259739	...	0.035377	-0.255434	1.382397
38	0.180435	0.864527	...	2.299535	3.247656	1.748673
39	-0.589421	-0.321110	...	0.035377	0.912263	-0.023631

	6RF_VPP_MAX	6RF_VPP_MIN	6TH_VALVE_AVG	6TH_VALVE_MAX	6TH_VALVE_MIN	\
0	1.360938	1.409973	-0.210652	-0.228309	-0.078011	
1	1.795734	0.984849	0.714565	0.446081	0.546076	
2	-0.178472	0.145787	-2.325434	-1.015098	-1.700636	
3	-0.613267	-0.480712	-0.475000	-0.677903	-0.202828	
4	0.702870	-0.055588	0.714565	-0.115911	0.920528	
5	1.360938	1.823910	0.714565	4.267626	-1.638227	
6	0.926143	1.409973	-1.003695	0.108886	-1.700636	
7	0.479597	0.358349	1.111087	0.558479	0.982936	
8	-0.836540	-0.480712	0.318043	0.108886	0.421258	
9	-1.059813	-1.107212	-1.003695	-0.790301	-0.265237	
10	-2.164428	-1.733711	-0.078478	-0.340708	0.171624	
11	-0.613267	-1.107212	1.111087	0.333683	1.232571	
12	0.479597	1.197411	0.582391	0.895675	-0.078011	
13	0.044802	-0.480712	0.714565	-0.565504	1.482205	
14	0.044802	-0.268150	-1.664565	-0.565504	-1.451001	
15	-0.401745	-0.480712	1.639782	1.682463	0.920528	
16	-1.283087	-1.107212	-0.342826	-0.453106	0.171624	
17	-1.494609	-1.733711	-0.475000	-1.239895	-0.140419	
18	1.137665	0.984849	-0.210652	-0.340708	-0.327645	
19	0.044802	-0.055588	0.846739	1.120471	0.608484	
20	2.242280	2.249035	0.318043	-0.340708	0.546076	
21	-0.401745	-0.480712	-1.796738	-0.790301	-1.575818	
22	0.479597	0.358349	0.582391	-0.003512	0.670893	

23	0.479597	0.570911	1.243260	0.333683	1.045345
24	0.702870	0.783474	0.318043	-0.115911	0.358850
25	-0.178472	-0.055588	0.582391	0.108886	1.045345
26	0.702870	0.783474	0.714565	-0.003512	0.795710
27	-1.059813	-1.107212	0.714565	0.108886	0.795710
28	-1.059813	-1.319774	-0.078478	-0.565504	-0.202828
29	-1.283087	-0.682087	-1.928912	-1.464691	-1.763044
30	-0.178472	-0.055588	-1.532391	-0.228309	-1.950270
31	-0.178472	-0.268150	0.185869	-0.228309	0.358850
32	-0.401745	-0.894649	0.846739	0.108886	0.608484
33	-0.613267	-0.894649	1.111087	0.783276	0.546076
34	1.360938	0.984849	0.846739	-0.453106	1.294979
35	0.702870	0.570911	-0.475000	-0.115911	-0.202828
36	1.795734	1.823910	-0.342826	1.008073	-1.201366
37	1.360938	1.409973	-0.078478	-0.003512	-0.140419
38	1.795734	1.622535	0.318043	-0.228309	0.421258
39	-0.178472	0.145787	0.053696	-0.115911	0.234032

6WFS\_PM

0	-1.533705
1	-1.066290
2	-0.131460
3	0.335954
4	-0.131460
5	-1.533705
6	-1.066290
7	-0.598875
8	-0.131460
9	0.335954
10	1.738199
11	1.738199
12	-1.533705
13	0.335954
14	0.335954
15	0.803369
16	1.270784
17	1.738199
18	-1.533705
19	-0.598875
20	-1.066290
21	0.335954
22	0.803369
23	1.270784
24	-0.598875
25	0.335954
26	-1.066290
27	0.803369



```

28  1.270784
29 -0.131460
30 -0.598875
31 -0.131460
32  0.803369
33  1.270784
34 -0.598875
35 -1.066290
36 -1.533705
37 -1.066290
38 -1.533705
39  1.738199

```

[40 rows x 198 columns]

```
[132]: Arcing_Issue_standard.to_csv('Arcing_Issue_standard_195.csv')
```

### 1.3 train data with RF

```
[ ]: from sklearn.ensemble import RandomForestClassifier

rnd_clf = RandomForestClassifier(n_estimators=100, min_samples_split=10,
    ↪min_samples_leaf=3, max_features=143, n_jobs=-1, random_state=42)
rnd_clf.fit(train_data, train_labels)
test_pred = rnd_clf.predict(test_data)
```

#### 1.3.1 accuracy

```
[134]: from sklearn.metrics import accuracy_score
accuracy_score(test_labels, test_pred)
```

```
[134]: 0.75
```

#### 1.3.2 feature importances

```
[135]: importances_0=rnd_clf.feature_importances_
```

```
[136]: cols_0=data.columns[2:-1]
df_importances_0=pd.DataFrame({'Variable':cols_0,'Importance':importances_0})
df_importances_0=df_importances_0.sort_values(by='Importance',ascending=False).
    ↪head(20)
df_importances_0
```

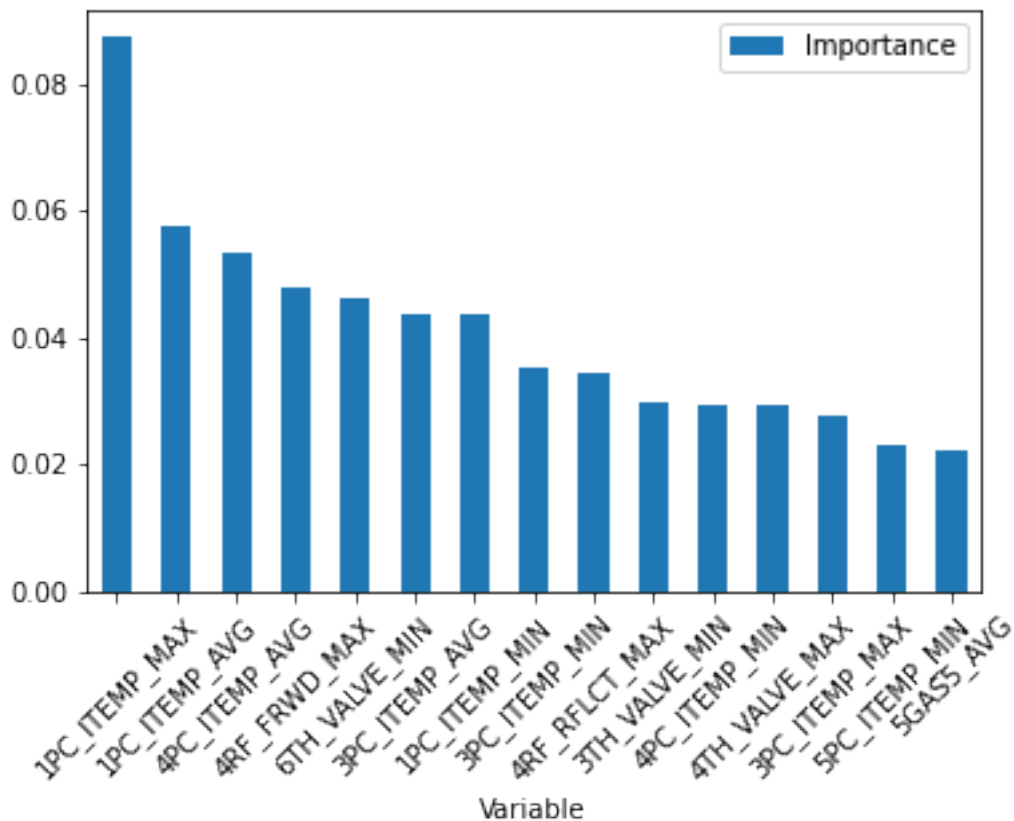
```
[136]:
```

	Variable	Importance
1	1PC_ITEMP_MAX	0.087206
0	1PC_ITEMP_AVG	0.057342

75	4PC_ITEMP_AVG	0.053386
94	4RF_FRWD_MAX	0.047932
193	6TH_VALVE_MIN	0.046300
44	3PC_ITEMP_AVG	0.043738
2	1PC_ITEMP_MIN	0.043642
46	3PC_ITEMP_MIN	0.035109
100	4RF_RFLCT_MAX	0.034225
64	3TH_VALVE_MIN	0.029806
77	4PC_ITEMP_MIN	0.029430
109	4TH_VALVE_MAX	0.029139
45	3PC_ITEMP_MAX	0.027485
123	5PC_ITEMP_MIN	0.023096
118	5GAS5_AVG	0.022334
167	6PRESSURE_AVG	0.021672
168	6PRESSURE_MAX	0.021636
144	5RF_LOAD_MIN	0.021457
62	3TH_VALVE_AVG	0.019791
36	2RF_VDC_MIN	0.019090

```
[137]: df_importances_0.head(15).plot(kind='bar',x='Variable', y='Importance', rot=45)
```

```
[137]: <matplotlib.axes._subplots.AxesSubplot at 0x1bb27d6f888>
```



```
[138]: Arcing_Issue_standard.loc[:, df_importances_0.iloc[:15,0]].head()
```

```
[138]:
```

	1PC_ITEMP_MAX	1PC_ITEMP_AVG	4PC_ITEMP_AVG	4RF_FRWD_MAX	6TH_VALVE_MIN \
0	-0.284609	-0.384757	0.990137	-0.477036	-0.078011
1	1.471051	1.435313	0.751010	1.000237	0.546076
2	0.812679	0.792935	1.109701	-0.723248	-1.700636
3	-0.723524	-0.705946	-1.162010	-0.230824	-0.202828
4	0.812679	0.899998	0.870573	1.000237	0.920528

	3PC_ITEMP_AVG	1PC_ITEMP_MIN	3PC_ITEMP_MIN	4RF_RFLCT_MAX	3TH_VALVE_MIN \
0	0.131372	-0.399929	-0.221346	-0.797009	-0.26968
1	1.086804	1.428320	1.066484	1.420755	0.26968
2	1.086804	0.818903	1.066484	-1.536264	1.34840
3	-1.015147	-0.806207	-0.773273	-0.797009	1.34840
4	0.991261	0.818903	1.066484	0.681501	-1.34840

	4PC_ITEMP_MIN	4TH_VALVE_MAX	3PC_ITEMP_MAX	5PC_ITEMP_MIN	5GAS5_AVG
0	0.792619	-1.664790	0.561551	1.039797	-1.068525
1	0.792619	0.418523	1.167268	0.643684	-0.031560
2	1.023199	-0.871147	1.066315	1.039797	-0.130748
3	-1.282602	-1.367174	-1.255602	-1.072806	0.752927
4	0.907909	-1.069558	0.965362	0.775721	0.013526

## 1.4 GridSearchCV

```
[139]: '''
RandomForestClassifier(
    n_estimators='warn',
    criterion='gini',
    max_depth=None,
    min_samples_split=2,
    min_samples_leaf=1,
    min_weight_fraction_leaf=0.0,
    max_features='auto',
    max_leaf_nodes=None,
    min_impurity_decrease=0.0,
    min_impurity_split=None,
    bootstrap=True,
    oob_score=False,
    n_jobs=None,
    random_state=None,
    verbose=0,
    warm_start=False,
    class_weight=None,
```

```
)  
'''
```

```
[139]: "\nRandomForestClassifier(\n    n_estimators='warn',\n    criterion='gini',\n    max_depth=None,\n    min_samples_split=2,\n    min_samples_leaf=1,\n    min_weight_fraction_leaf=0.0,\n    max_features='auto',\n    max_leaf_nodes=None,\n    min_impurity_decrease=0.0,\n    min_impurity_split=None,\n    bootstrap=True,\n    oob_score=False,\n    n_jobs=None,\n    random_state=None,\n    verbose=0,\n    warm_start=False,\n    class_weight=None,\n)\n"
```

```
rnd_clf = RandomForestClassifier(n_estimators=100, min_samples_split=10,  
min_samples_leaf=3, max_features=143, n_jobs=-1, random_state=42)
```

0.75

```
[140]: from sklearn.ensemble import RandomForestClassifier  
  
rnd_clf_grid = RandomForestClassifier(n_jobs=-1, random_state=42)
```

```
[ ]: from sklearn.model_selection import GridSearchCV  
  
param_distributions = {"n_estimators": [100,200], "max_features":  
    ↳ [120,130], "min_samples_split": [3,5], "min_samples_leaf": [1,3,5,7]}  
grid_search_cv = GridSearchCV(rnd_clf_grid, param_distributions, cv=4, ↳  
    ↳ verbose=2, scoring='accuracy') # 4-fold  
grid_search_cv.fit(train_data, train_labels)
```

```
[142]: grid_search_cv_best=grid_search_cv.best_estimator_  
grid_search_cv_best
```

```
[142]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',  
max_depth=None, max_features=120, max_leaf_nodes=None,  
min_impurity_decrease=0.0, min_impurity_split=None,  
min_samples_leaf=7, min_samples_split=3,  
min_weight_fraction_leaf=0.0, n_estimators=100,  
n_jobs=-1, oob_score=False, random_state=42, verbose=0,  
warm_start=False)
```

```
[143]: grid_search_cv.best_score_
```

[143]: 0.75

```
[ ]: grid_search_cv_best.fit(train_data, train_labels)
```

```
[145]: train_pred = grid_search_cv_best.predict(train_data)  
accuracy_score(train_labels, train_pred)
```

```
C:\Users\pj007\Anaconda3\lib\site-packages\sklearn\ensemble\base.py:158:
DeprecationWarning: `np.int` is a deprecated alias for the builtin `int`. To
silence this warning, use `int` by itself. Doing this will not modify any
behavior and is safe. When replacing `np.int`, you may wish to use e.g.
`np.int64` or `np.int32` to specify the precision. If you wish to review your
current use, check the release note link for additional information.
Deprecated in NumPy 1.20; for more details and guidance:
https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
dtype=np.int)
```

```
[145]: 0.84375
```

## 1.5 use RF model to predict labels of test set

```
[146]: test_pred = grid_search_cv_best.predict(test_data)
test_pred
```

```
C:\Users\pj007\Anaconda3\lib\site-packages\sklearn\ensemble\base.py:158:
DeprecationWarning: `np.int` is a deprecated alias for the builtin `int`. To
silence this warning, use `int` by itself. Doing this will not modify any
behavior and is safe. When replacing `np.int`, you may wish to use e.g.
`np.int64` or `np.int32` to specify the precision. If you wish to review your
current use, check the release note link for additional information.
Deprecated in NumPy 1.20; for more details and guidance:
https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
dtype=np.int)
```

```
[146]: array(['N', 'Y', 'Y', 'N', 'Y', 'N', 'Y', 'Y'], dtype=object)
```

## 1.6 model evaluation

### 1.6.1 classification\_report

```
[147]: df=pd.DataFrame({'test_label':test_labels,'test_pred':test_pred})
df
```

```
[147]:   test_label test_pred
32         N         N
33         Y         Y
34         Y         Y
35         Y         N
36         Y         Y
37         N         N
38         Y         Y
39         Y         Y
```

```
[148]: import sklearn.metrics as sm
cr = sm.classification_report(test_labels,test_pred)
print(cr)
```

	precision	recall	f1-score	support
N	0.67	1.00	0.80	2
Y	1.00	0.83	0.91	6
accuracy			0.88	8
macro avg	0.83	0.92	0.85	8
weighted avg	0.92	0.88	0.88	8

### 1.6.2 confusion matrix

```
[149]: from sklearn.metrics import confusion_matrix
confusion_matrix(test_labels,test_pred)
```

```
[149]: array([[2, 0],
             [1, 5]], dtype=int64)
```

```
[150]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import KFold, cross_val_score
from sklearn.metrics import confusion_matrix, recall_score,
↳ classification_report
```

```
[151]: def plot_confusion_matrix(cm, classes,normalize=False,title='Confusion_
↳ matrix',cmap=plt.cm.Blues):
    """
    This function prints and plots the confusion matrix.
    Normalization can be applied by setting `normalize=True`.
    """
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')
    print(cm)
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick_marks = np.arange(len(classes))
    plt.xticks(tick_marks, classes, rotation=45)
```

```

plt.yticks(tick_marks, classes)
fmt = '.2f' if normalize else 'd'
thresh = cm.max() / 2.
for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
    plt.text(j, i, format(cm[i, j], fmt),
             horizontalalignment="center",
             color="white" if cm[i, j] > thresh else "black")
plt.tight_layout()
plt.ylabel('True label')
plt.xlabel('Predicted label')

```

```

[152]: import itertools
cnf_matrix = confusion_matrix(test_labels, test_pred) #
class_names = [0,1]
plt.figure()
plot_confusion_matrix(cnf_matrix, classes = class_names, title = 'Confusion_
↪matrix') #
np.set_printoptions(precision=2)
print('Accuracy:', (cnf_matrix[1,1]+cnf_matrix[0,0])/
↪(cnf_matrix[1,1]+cnf_matrix[0,1]+cnf_matrix[0,0]+cnf_matrix[1,0]))
print('Recall:', cnf_matrix[1,1]/(cnf_matrix[1,1]+cnf_matrix[1,0]))
print('Precision:', cnf_matrix[1,1]/(cnf_matrix[1,1]+cnf_matrix[0,1]))
print('Specificity:', cnf_matrix[0,0]/(cnf_matrix[0,1]+cnf_matrix[0,0]))
plt.show()

```

Confusion matrix, without normalization

```
[[2 0]
```

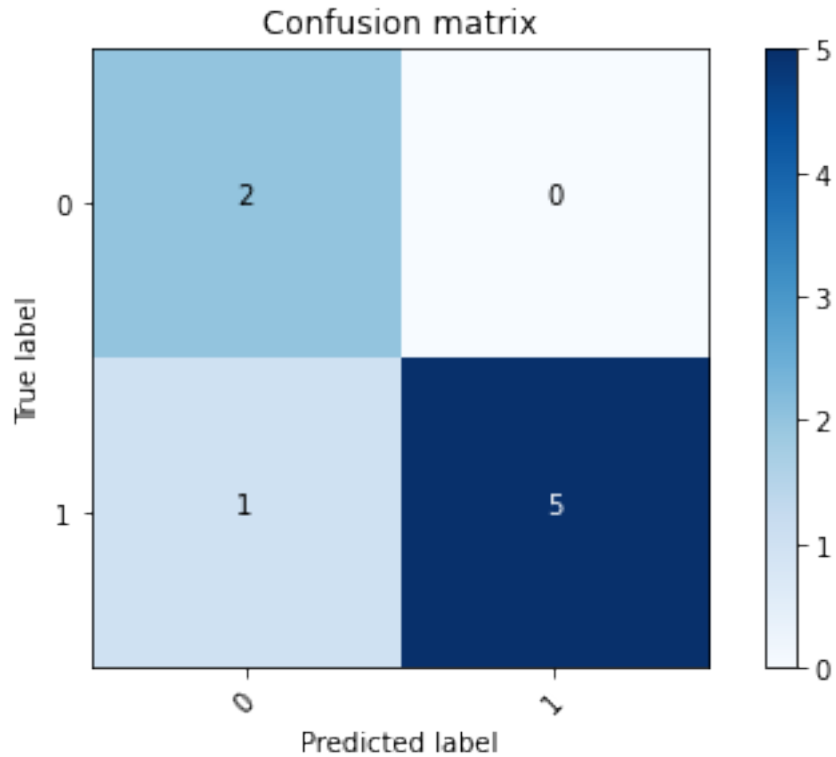
```
 [1 5]]
```

Accuracy: 0.875

Recall: 0.8333333333333334

Precision: 1.0

Specificity: 1.0



### 1.6.3 ROC curve

```
[153]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import sklearn.metrics as metrics

def plot_ROC(labels,preds):
    """
    Args:
        labels : ground truth
        preds : model prediction
        savepath : save path
    """
    fpr1, tpr1, threshold1 = metrics.roc_curve(labels, preds) #

    roc_auc1 = metrics.auc(fpr1, tpr1) # auc auc
    plt.figure()
    lw = 2
    plt.figure(figsize=(10, 10))
    plt.plot(fpr1, tpr1, color='darkorange',
             lw=lw, label='AUC = %0.2f' % roc_auc1) #
```



```
plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
plt.xlim([-0.05, 1.05])
plt.ylim([-0.05, 1.05])
plt.xlabel('1 - Specificity')
plt.ylabel('Sensitivity')
# plt.title('ROCs for Densenet')
plt.legend(loc="lower right")
# plt.show()
```

```
[154]: test_labels_1=np.array(test_labels)
test_labels_1=[1 if x=='Y' else 0 for x in test_labels_1 ]
test_labels_1
```

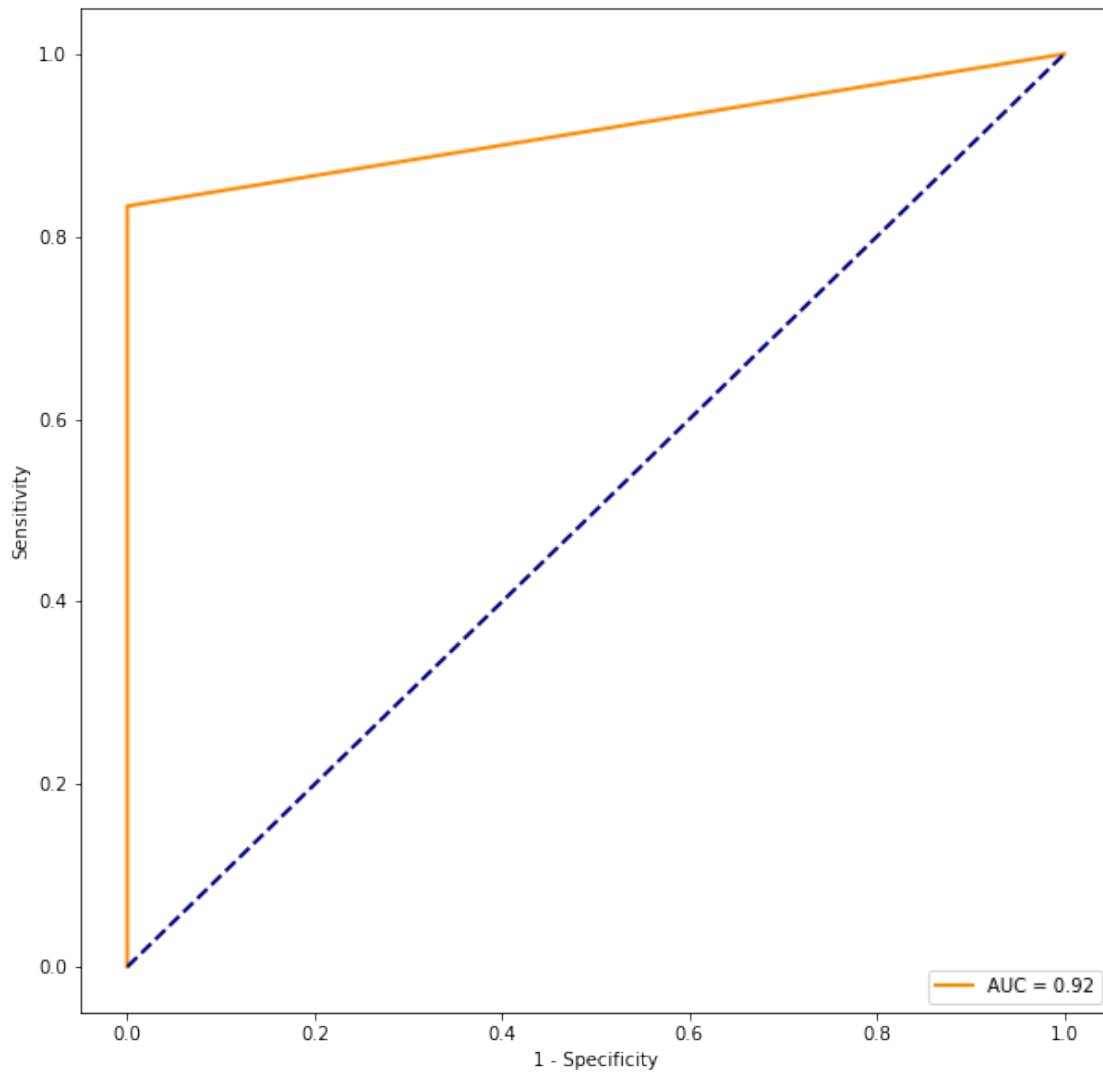
```
[154]: [0, 1, 1, 1, 1, 0, 1, 1]
```

```
[155]: test_pred_1=np.array(test_pred)
test_pred_1=[1 if x=='Y' else 0 for x in test_pred_1 ]
test_pred_1
```

```
[155]: [0, 1, 1, 0, 1, 0, 1, 1]
```

```
[156]: plot_ROC(test_labels_1,test_pred_1)
```

<Figure size 432x288 with 0 Axes>



#### 1.6.4 feature importances

```
[157]: importances=grid_search_cv_best.feature_importances_
```

```
[158]: cols=data.columns[2:-1]
df_importances_1=pd.DataFrame({'Variable':cols,'Importance':importances})
df_importances_1=df_importances_1.sort_values(by='Importance',ascending=False).
↳head(20)
df_importances_1
```

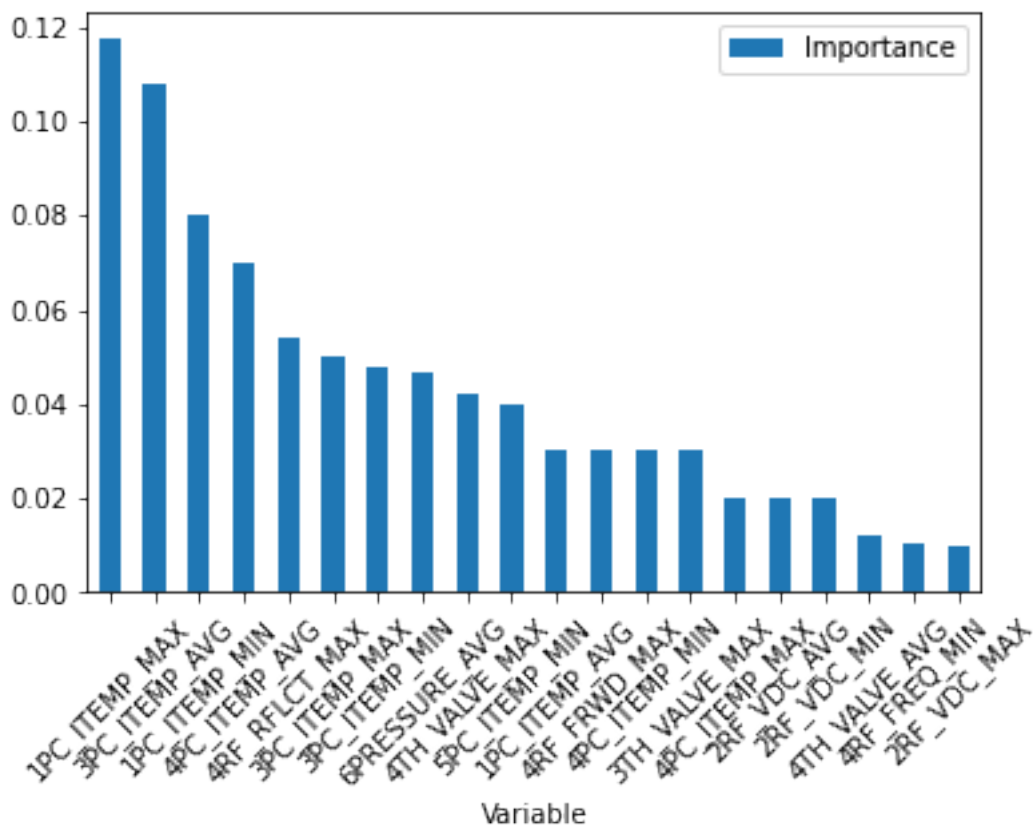
```
[158]:
```

	Variable	Importance
1	1PC_ITEMP_MAX	0.117355
44	3PC_ITEMP_AVG	0.108142

2	1PC_ITEMP_MIN	0.080000
75	4PC_ITEMP_AVG	0.070000
100	4RF_RFLCT_MAX	0.054301
45	3PC_ITEMP_MAX	0.050000
46	3PC_ITEMP_MIN	0.047715
167	6PRESSURE_AVG	0.046483
109	4TH_VALVE_MAX	0.042429
123	5PC_ITEMP_MIN	0.039800
0	1PC_ITEMP_AVG	0.030200
94	4RF_FRWD_MAX	0.030000
77	4PC_ITEMP_MIN	0.030000
63	3TH_VALVE_MAX	0.030000
76	4PC_ITEMP_MAX	0.020000
34	2RF_VDC_AVG	0.020000
36	2RF_VDC_MIN	0.020000
108	4TH_VALVE_AVG	0.012285
92	4RF_FREQ_MIN	0.010164
35	2RF_VDC_MAX	0.010000

```
[159]: df_importances_1.head(20).plot(kind='bar',x='Variable', y='Importance', rot=45)
```

```
[159]: <matplotlib.axes._subplots.AxesSubplot at 0x1bb269dca48>
```



```
[160]: Arcing_Issue_standard.loc[:, df_importances_1.iloc[:20,0]].head()
```

```
[160]: 1PC_ITEMP_MAX 3PC_ITEMP_AVG 1PC_ITEMP_MIN 4PC_ITEMP_AVG 4RF_RFLCT_MAX \
0      -0.284609      0.131372     -0.399929      0.990137     -0.797009
1       1.471051      1.086804      1.428320      0.751010      1.420755
2       0.812679      1.086804      0.818903      1.109701     -1.536264
3      -0.723524     -1.015147     -0.806207     -1.162010     -0.797009
4       0.812679      0.991261      0.818903      0.870573      0.681501

      3PC_ITEMP_MAX 3PC_ITEMP_MIN 6PRESSURE_AVG 4TH_VALVE_MAX 5PC_ITEMP_MIN \
0       0.561551     -0.221346      0.081557     -1.664790      1.039797
1       1.167268      1.066484     -0.676131      0.418523      0.643684
2       1.066315      1.066484      1.091807     -0.871147      1.039797
3      -1.255602     -0.773273     -0.676131     -1.367174     -1.072806
4       0.965362      1.066484     -0.339381     -1.069558      0.775721

      1PC_ITEMP_AVG 4RF_FRWD_MAX 4PC_ITEMP_MIN 3TH_VALVE_MAX 4PC_ITEMP_MAX \
0      -0.384757     -0.477036      0.792619     -0.455407      0.914145
1       1.435313      1.000237      0.792619     -0.455407      0.792765
2       0.792935     -0.723248      1.023199     -0.102122      1.156906
3      -0.705946     -0.230824     -1.282602     -0.190443     -1.027939
4       0.899998      1.000237      0.907909     -0.720371      0.792765

      2RF_VDC_AVG 2RF_VDC_MIN 4TH_VALVE_AVG 4RF_FREQ_MIN 2RF_VDC_MAX
0       1.044185      0.254219     -0.578877      1.264786      1.539865
1      -0.240966      0.254219     -0.167231      1.632723     -0.778991
2      -1.526117     -1.623089     -0.270143      0.160973     -0.778991
3       0.401610      0.254219     -0.270143      0.528910      0.380437
4      -1.526117     -0.997320     -0.784700     -0.758871     -1.358705
```

## 1.7 save model using joblib

```
[161]: my_model =grid_search_cv_best
my_model
```

```
[161]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                             max_depth=None, max_features=120, max_leaf_nodes=None,
                             min_impurity_decrease=0.0, min_impurity_split=None,
                             min_samples_leaf=7, min_samples_split=3,
                             min_weight_fraction_leaf=0.0, n_estimators=100,
                             n_jobs=-1, oob_score=False, random_state=42, verbose=0,
                             warm_start=False)
```

```
[162]: #save my_model
import joblib
#from sklearn.externals import joblib
joblib.dump(my_model, "random_forest.pkl")
```

```
[162]: ['random_forest.pkl']
```