

Notebook

November 9, 2025

```
[1]: import pandas as pd  
import scipy.stats as stats
```

```
[2]: df = pd.read_csv("data/original-data.csv")  
df.rename(columns={'V1': 'X_Minimum'}, inplace=True)  
df.rename(columns={'V2': 'X_Maximum'}, inplace=True)  
df.rename(columns={'V3': 'Y_Minimum'}, inplace=True)  
df.rename(columns={'V4': 'Y_Maximum'}, inplace=True)  
df.rename(columns={'V5': 'Pixels_Areas'}, inplace=True)  
df.rename(columns={'V6': 'X_Perimeter'}, inplace=True)  
df.rename(columns={'V7': 'Y_Perimeter'}, inplace=True)  
df.rename(columns={'V8': 'Sum_of_Luminosity'}, inplace=True)  
df.rename(columns={'V9': 'Minimum_of_Luminosity'}, inplace=True)  
df.rename(columns={'V10': 'Maximum_of_Luminosity'}, inplace=True)  
df.rename(columns={'V11': 'Length_of_Conveyer'}, inplace=True)  
df.rename(columns={'V12': 'TypesOfSteel_A300'}, inplace=True)  
df.rename(columns={'V13': 'TypesOfSteel_A400'}, inplace=True)  
df.rename(columns={'V14': 'Steel_Plate_Thickness'}, inplace=True)  
df.rename(columns={'V15': 'Edges_Index'}, inplace=True)  
df.rename(columns={'V16': 'Empty_Index'}, inplace=True)  
df.rename(columns={'V17': 'Square_Index'}, inplace=True)  
df.rename(columns={'V18': 'Outside_X_Index'}, inplace=True)  
df.rename(columns={'V19': 'Edges_X_Index'}, inplace=True)  
df.rename(columns={'V20': 'Edges_Y_Index'}, inplace=True)  
df.rename(columns={'V21': 'Outside_Global_Index'}, inplace=True)  
df.rename(columns={'V22': 'LogOfAreas'}, inplace=True)  
df.rename(columns={'V23': 'Log_X_Index'}, inplace=True)  
df.rename(columns={'V24': 'Log_Y_Index'}, inplace=True)  
df.rename(columns={'V25': 'Orientation_Index'}, inplace=True)  
df.rename(columns={'V26': 'Luminosity_Index'}, inplace=True)  
df.rename(columns={'V27': 'SigmoidOfAreas'}, inplace=True)  
df.rename(columns={'V28': 'Pastry'}, inplace=True)  
df.rename(columns={'V29': 'Z_Scratch'}, inplace=True)  
df.rename(columns={'V30': 'K_Scratch'}, inplace=True)  
df.rename(columns={'V31': 'Stains'}, inplace=True)  
df.rename(columns={'V32': 'Dirtiness'}, inplace=True)  
df.rename(columns={'V33': 'Bumps'}, inplace=True)  
df.rename(columns={'Class': 'Class'}, inplace=True) # Other_Faults
```

df

[2]:	X_Minimum	X_Maximum	Y_Minimum	Y_Maximum	Pixels_Areas	X_Perimeter	\	
0	42	50	270900	270944	267	17		
1	645	651	2538079	2538108	108	10		
2	829	835	1553913	1553931	71	8		
3	853	860	369370	369415	176	13		
4	1289	1306	498078	498335	2409	60		
...		
1936	249	277	325780	325796	273	54		
1937	144	175	340581	340598	287	44		
1938	145	174	386779	386794	292	40		
1939	137	170	422497	422528	419	97		
1940	1261	1281	87951	87967	103	26		
	Y_Perimeter	Sum_of_Luminosity	Minimum_of_Luminosity	\				
0	44	24220	76					
1	30	11397	84					
2	19	7972	99					
3	45	18996	99					
4	260	246930	37					
...				
1936	22	35033	119					
1937	24	34599	112					
1938	22	37572	120					
1939	47	52715	117					
1940	22	11682	101					
	Maximum_of_Luminosity	...	Orientation_Index	Luminosity_Index	\			
0	108	...	0.8182	-0.2913				
1	123	...	0.7931	-0.1756				
2	125	...	0.6667	-0.1228				
3	126	...	0.8444	-0.1568				
4	126	...	0.9338	-0.1992				
...				
1936	141	...	-0.4286	0.0026				
1937	133	...	-0.4516	-0.0582				
1938	140	...	-0.4828	0.0052				
1939	140	...	-0.0606	-0.0171				
1940	133	...	-0.2000	-0.1139				
	SigmoidOfAreas	Pastry	Z_Scratch	K_Scratch	Stains	Dirtiness	Bumps	\
0	0.5822	1	0	0	0	0	0	
1	0.2984	1	0	0	0	0	0	
2	0.2150	1	0	0	0	0	0	
3	0.5212	1	0	0	0	0	0	
4	1.0000	1	0	0	0	0	0	

```

...
1936      0.7254    0    0    0    0    0    0    0
1937      0.8173    0    0    0    0    0    0    0
1938      0.7079    0    0    0    0    0    0    0
1939      0.9919    0    0    0    0    0    0    0
1940      0.5296    0    0    0    0    0    0    0

      Class
0      1
1      1
2      1
3      1
4      1
...
1936      2
1937      2
1938      2
1939      2
1940      2

```

[1941 rows x 34 columns]

[3]: df.duplicated().any()

[3]: np.False_

No duplicates

1 Z-Score outlier detection

```

[4]: # All Columns except the ones we're predicting ('Pastry', 'Z_Scratch', □
      ↵ 'K_Scratch', 'Stains', 'Dirtiness', 'Bumps', 'Other_Faults_(Class)')
columns_to_check_outliers = [
    'X_Minimum', 'X_Maximum', 'Y_Minimum', 'Y_Maximum', 'Pixels_Areas', □
    ↵ 'X_Perimeter',
    'Y_Perimeter', 'Sum_of_Luminosity', 'Minimum_of_Luminosity', □
    ↵ 'Maximum_of_Luminosity',
    'Length_of_Conveyer', 'TypesOfSteel_A300', 'TypesOfSteel_A400', □
    ↵ 'Steel_Plate_Thickness',
    'Edges_Index', 'Empty_Index', 'Square_Index', 'Outside_X_Index', □
    ↵ 'Edges_X_Index',
    'Edges_Y_Index', 'Outside_Global_Index', 'LogOfAreas', 'Log_X_Index', □
    ↵ 'Log_Y_Index',
    'Orientation_Index', 'Luminosity_Index', 'SigmoidOfAreas'
]
# Ignoring certain columns without outliers - to use a more aggressive outlier □
      ↵ detection on the others

```

```

columns_to_ignore = [
    'X_Minimum', 'X_Maximum', 'Length_of_Conveyer', 'TypesOfSteel_A300',
    'TypesOfSteel_A400',
    'Edges_Index', 'Square_Index', 'Edges_X_Index', 'Edges_Y_Index',
    'Outside_Global_Index',
    'Orientation_Index', 'SigmoidOfAreas'
]
columns_to_check_outliers = [col for col in columns_to_check_outliers if col
    not in columns_to_ignore]
print(columns_to_check_outliers)
z_score = stats.zscore(df[columns_to_check_outliers])
df_clean = df.copy()
df_clean = df_clean[(abs(z_score) < 4).all(axis=1)]
['Y_Minimum', 'Y_Maximum', 'Pixels_Areas', 'X_Perimeter', 'Y_Perimeter',
'Sum_of_Luminosity', 'Minimum_of_Luminosity', 'Maximum_of_Luminosity',
'Steel_Plate_Thickness', 'Empty_Index', 'Outside_X_Index', 'LogOfAreas',
'Log_X_Index', 'Log_Y_Index', 'Luminosity_Index']

```

2 Modifikovaná 5% 95% Metóda

```

[5]: # All Columns except the ones we're predicting ('Pastry', 'Z_Scratch',
      'K_Scratch', 'Stains', 'Dirtiness', 'Bumps', 'Other_Faults_(Class)')
columns_to_check_outliers = [
    'X_Minimum', 'X_Maximum', 'Y_Minimum', 'Y_Maximum', 'Pixels_Areas',
    'X_Perimeter',
    'Y_Perimeter', 'Sum_of_Luminosity', 'Minimum_of_Luminosity',
    'Maximum_of_Luminosity',
    'TypesOfSteel_A300', 'TypesOfSteel_A400', 'Steel_Plate_Thickness',
    'Edges_Index',
    'Empty_Index', 'Square_Index', 'Outside_X_Index', 'Edges_X_Index',
    'Edges_Y_Index',
    'Outside_Global_Index', 'LogOfAreas', 'Log_X_Index', 'Log_Y_Index',
    'Orientation_Index',
    'Luminosity_Index', 'SigmoidOfAreas'
]
# Ignoring certain columns without outliers - to use a more aggressive outlier
# detection on the others
columns_to_ignore = [
    'X_Minimum', 'X_Maximum', 'Length_of_Conveyer', 'TypesOfSteel_A300',
    'TypesOfSteel_A400',
    'Edges_Index', 'Square_Index', 'Edges_X_Index', 'Edges_Y_Index',
    'Outside_Global_Index',
    'Orientation_Index', 'SigmoidOfAreas',
    # Ignoring columns that were well handled by Z-Score already or don't
    # require more cleaning

```

```

    'Y_Minimum', 'Y_Maximum', 'Minimum_of_Luminosity', 'Maximum_of_Luminosity',
    ↵'Steel_Plate_Thickness', 'Empty_Index',
    'Log_X_Index', 'Log_Y_Index', 'Luminosity_Index'
]
columns_to_check_outliers = [col for col in columns_to_check_outliers if col
    ↵not in columns_to_ignore]
print(columns_to_check_outliers)

shaving_ranges = [
    (0.0, 0.98), (0.0, 0.98), (0.0, 0.98), (0.0, 0.98), (0.0, 0.98), (0.0, 0.98)
]

for idx, col in enumerate(columns_to_check_outliers):
    lower = 0
    higher = 1
    df_CO_lower = df_clean[col].quantile(shaving_ranges[idx][lower])
    df_CO_upper = df_clean[col].quantile(shaving_ranges[idx][higher])

    df_clean[col] = df[col].where(
        (df_clean[col] >= df_CO_lower) & (df_clean[col] <= df_CO_upper)
    )
    df_clean.dropna(inplace = True)

```

['Pixels_Areas', 'X_Perimeter', 'Y_Perimeter', 'Sum_of_Luminosity',
 'Outside_X_Index', 'LogOfAreas']

[6]: df_clean = df_clean.drop(columns=["Y_Minimum", "Y_Maximum", "Edges_Index",
 ↵"Empty_Index"], errors="ignore")
 df_clean.describe()

	X_Minimum	X_Maximum	Pixels_Areas	X_Perimeter	Y_Perimeter	\
count	1613.000000	1613.000000	1613.000000	1613.000000	1613.000000	
mean	629.401116	655.522009	632.513329	48.324241	39.787353	
std	517.427471	507.119546	1334.099392	64.058738	47.378852	
min	0.000000	4.000000	2.000000	2.000000	1.000000	
25%	114.000000	181.000000	77.000000	14.000000	12.000000	
50%	563.000000	583.000000	141.000000	22.000000	21.000000	
75%	1090.000000	1106.000000	349.000000	47.000000	42.000000	
max	1705.000000	1713.000000	6277.000000	405.000000	330.000000	
	Sum_of_Luminosity	Minimum_of_Luminosity	Maximum_of_Luminosity	\		
count	1613.000000	1613.000000	1613.000000			
mean	66874.983881	90.650961	129.055797			
std	139509.821414	27.155015	16.596791			
min	250.000000	0.000000	70.000000			
25%	8602.000000	77.000000	124.000000			
50%	16182.000000	95.000000	127.000000			

75%	37460.000000	109.000000	135.000000		
max	652005.000000	179.000000	199.000000		
	Length_of_Conveyer	TypesOfSteel_A300	...	Orientation_Index \	
count	1613.000000	1613.000000	...	1613.000000	
mean	1468.679479	0.456293	...	0.121622	
std	150.098130	0.498240	...	0.495092	
min	1227.000000	0.000000	...	-0.970600	
25%	1358.000000	0.000000	...	-0.250000	
50%	1364.000000	0.000000	...	0.153900	
75%	1656.000000	1.000000	...	0.545400	
max	1794.000000	1.000000	...	0.946700	
	Luminosity_Index	SigmoidOfAreas	Pastry	Z_Scratch \	
count	1613.000000	1613.000000	1613.000000	1613.000000	
mean	-0.131680	0.524665	0.091754	0.115313	
std	0.140006	0.321606	0.288769	0.319498	
min	-0.609600	0.119000	0.000000	0.000000	
25%	-0.199500	0.230000	0.000000	0.000000	
50%	-0.132600	0.402500	0.000000	0.000000	
75%	-0.057500	0.897100	0.000000	0.000000	
max	0.457300	1.000000	1.000000	1.000000	
	K_Scratch	Stains	Dirtiness	Bumps	Class
count	1613.000000	1613.000000	1613.000000	1613.000000	1613.000000
mean	0.106634	0.044637	0.034098	0.243025	1.364538
std	0.308743	0.206570	0.181537	0.429043	0.481450
min	0.000000	0.000000	0.000000	0.000000	1.000000
25%	0.000000	0.000000	0.000000	0.000000	1.000000
50%	0.000000	0.000000	0.000000	0.000000	1.000000
75%	0.000000	0.000000	0.000000	0.000000	2.000000
max	1.000000	1.000000	1.000000	1.000000	2.000000

[8 rows x 30 columns]

```
[7]: feature_cols = [*df_clean.columns[:len(df_clean.columns)-7]]
feature_cols
```

```
[7]: ['X_Minimum',
 'X_Maximum',
 'Pixels_Areas',
 'X_Perimeter',
 'Y_Perimeter',
 'Sum_of_Luminosity',
 'Minimum_of_Luminosity',
 'Maximum_of_Luminosity',
 'Length_of_Conveyer',
```

```
'TypesOfSteel_A300',
'TypesOfSteel_A400',
'Steel_Plate_Thickness',
'Square_Index',
'Outside_X_Index',
'Edges_X_Index',
'Edges_Y_Index',
'Outside_Global_Index',
'LogOfAreas',
'Log_X_Index',
'Log_Y_Index',
'Orientation_Index',
'Luminosity_Index',
'SigmoidOfAreas']
```

[8]: predict_cols = ["Pastry", "Z_Scratch", "K_Scratch", "K_Scratch", "Stains",
↳ "Dirtiness", "Bumps", "Class"]

[9]: df_clean["Class"].nunique()

[9]: 2

[10]: df_clean

	X_Minimum	X_Maximum	Pixels_Areas	X_Perimeter	Y_Perimeter	\
0	42	50	267.0	17.0	44.0	
1	645	651	108.0	10.0	30.0	
2	829	835	71.0	8.0	19.0	
3	853	860	176.0	13.0	45.0	
4	1289	1306	2409.0	60.0	260.0	
...	
1936	249	277	273.0	54.0	22.0	
1937	144	175	287.0	44.0	24.0	
1938	145	174	292.0	40.0	22.0	
1939	137	170	419.0	97.0	47.0	
1940	1261	1281	103.0	26.0	22.0	

	Sum_of_Luminosity	Minimum_of_Luminosity	Maximum_of_Luminosity	\
0	24220.0	76	108	
1	11397.0	84	123	
2	7972.0	99	125	
3	18996.0	99	126	
4	246930.0	37	126	
...	
1936	35033.0	119	141	
1937	34599.0	112	133	
1938	37572.0	120	140	

1939	52715.0	117	140				
1940	11682.0	101	133				
	Length_of_Conveyer	TypesOfSteel_A300	...	Orientation_Index	\		
0	1687	1	...	0.8182			
1	1687	1	...	0.7931			
2	1623	1	...	0.6667			
3	1353	0	...	0.8444			
4	1353	0	...	0.9338			
...			
1936	1360	0	...	-0.4286			
1937	1360	0	...	-0.4516			
1938	1360	0	...	-0.4828			
1939	1360	0	...	-0.0606			
1940	1360	1	...	-0.2000			
	Luminosity_Index	SigmoidOfAreas	Pastry	Z_Scratch	K_Scratch	Stains	\
0	-0.2913	0.5822	1	0	0	0	
1	-0.1756	0.2984	1	0	0	0	
2	-0.1228	0.2150	1	0	0	0	
3	-0.1568	0.5212	1	0	0	0	
4	-0.1992	1.0000	1	0	0	0	
...	
1936	0.0026	0.7254	0	0	0	0	
1937	-0.0582	0.8173	0	0	0	0	
1938	0.0052	0.7079	0	0	0	0	
1939	-0.0171	0.9919	0	0	0	0	
1940	-0.1139	0.5296	0	0	0	0	
	Dirtiness	Bumps	Class				
0	0	0	1				
1	0	0	1				
2	0	0	1				
3	0	0	1				
4	0	0	1				
...				
1936	0	0	2				
1937	0	0	2				
1938	0	0	2				
1939	0	0	2				
1940	0	0	2				

[1613 rows x 30 columns]

```
[11]: normalized_df = (df_clean[feature_cols] - df_clean[feature_cols].mean())/
    ↵df_clean[feature_cols].std()
```

```
[12]: normalized_df
```

```
[12]:      X_Minimum  X_Maximum  Pixels_Areas  X_Perimeter  Y_Perimeter  \
0      -1.135234  -1.194042   -0.273978   -0.488992    0.088914
1       0.030147  -0.008917   -0.393159   -0.598267   -0.206576
2       0.385752   0.353917   -0.420893   -0.629489   -0.438747
3       0.432136   0.403215   -0.342188   -0.551435    0.110021
4       1.274766   1.282692    1.331600    0.182266    4.647910
...
1936   -0.735178  -0.746416   -0.269480   0.088602   -0.375428
1937   -0.938105  -0.947552   -0.258986   -0.067504  -0.333215
1938   -0.936172  -0.949524   -0.255238   -0.129947  -0.375428
1939   -0.951633  -0.957411   -0.160043    0.759861    0.152233
1940   1.220652   1.233394   -0.396907   -0.348496  -0.375428

      Sum_of_Luminosity  Minimum_of_Luminosity  Maximum_of_Luminosity  \
0           -0.305749            -0.539531          -1.268667
1           -0.397664            -0.244926          -0.364878
2           -0.422214            0.307458          -0.244372
3           -0.343194            0.307458          -0.184120
4           1.290626            -1.975729          -0.184120
...
1936   -0.228242            1.043971          0.719669
1937   -0.231353            0.786191          0.237649
1938   -0.210042            1.080796          0.659417
1939   -0.101498            0.970319          0.659417
1940   -0.395621            0.381110          0.237649

      Length_of_Conveyer  TypesOfSteel_A300  ...  Outside_X_Index  \
0           1.454519            1.091255  ...  -0.514948
1           1.454519            1.091255  ...  -0.556364
2           1.028131            1.091255  ...  -0.552599
3           -0.770692            -0.915808  ...  -0.496123
4           -0.770692            -0.915808  ...  -0.217504
...
1936   -0.724056            -0.915808  ...  0.083706
1937   -0.724056            -0.915808  ...  0.166538
1938   -0.724056            -0.915808  ...  0.110061
1939   -0.724056            -0.915808  ...  0.223015
1940   -0.724056            1.091255  ...  -0.138436

      Edges_X_Index  Edges_Y_Index  Outside_Global_Index  LogOfAreas  \
0           -0.718842            0.695123            0.836722    0.237173
1           -0.159904            0.526690            0.836722   -0.417695
2            0.488015            0.429070            0.836722   -0.721056
3           -0.425551            0.695123            0.836722   -0.064356
4           -1.527877            0.636955            0.836722    1.828612
```

```

...
1936      -0.511940     -0.684202      ...      -1.263874      0.253332
1937       0.291911     -0.780305      ...      -1.263874      0.289482
1938       0.380028     -0.914342      ...      -1.263874      0.301977
1939      -1.282100     -1.026631      ...      -1.263874      0.563191
1940       0.570948     -0.684202      ...      -1.263874     -0.452012

    Log_X_Index  Log_Y_Index  Orientation_Index  Luminosity_Index \
0      -0.846876     0.865673      1.406965      -1.140097
1      -1.190733     0.399819      1.356268      -0.313702
2      -1.190733     -0.132917      1.100962      0.063426
3      -1.006553     0.890625      1.459885     -0.179421
4       0.054474     2.837130      1.640457     -0.482266
...
1936      0.651061     -0.264622      ...      -1.111353      0.959104
1937      0.772746     -0.196712      ...      -1.157808      0.524836
1938      0.692907     -0.336648      ...      -1.220827      0.977674
1939      0.847354     0.474417      ...      -0.368057      0.818395
1940      0.248564     -0.264622      ...      -0.649621      0.126995

    SigmoidOfAreas
0        0.178899
1      -0.703548
2      -0.962871
3      -0.010774
4       1.478005
...
1936      0.624165
1937      0.909918
1938      0.569750
1939      1.452819
1940      0.015345

[1613 rows x 23 columns]

```

```
[13]: df2 = normalized_df.join(df[predict_cols], how='inner', lsuffix='_caller', rsuffix='_other')
```

```
[14]: df2
```

```
[14]:   X_Minimum  X_Maximum  Pixels_Areas  X_Perimeter  Y_Perimeter \
0      -1.135234  -1.194042     -0.273978     -0.488992     0.088914
1       0.030147  -0.008917     -0.393159     -0.598267    -0.206576
2       0.385752   0.353917     -0.420893     -0.629489    -0.438747
3       0.432136   0.403215     -0.342188     -0.551435     0.110021
4      1.274766   1.282692      1.331600      0.182266     4.647910
...
...
```

1936	-0.735178	-0.746416	-0.269480	0.088602	-0.375428		
1937	-0.938105	-0.947552	-0.258986	-0.067504	-0.333215		
1938	-0.936172	-0.949524	-0.255238	-0.129947	-0.375428		
1939	-0.951633	-0.957411	-0.160043	0.759861	0.152233		
1940	1.220652	1.233394	-0.396907	-0.348496	-0.375428		
0	Sum_of_Luminosity	Minimum_of_Luminosity	Maximum_of_Luminosity	\\			
0	-0.305749	-0.539531	-1.268667				
1	-0.397664	-0.244926	-0.364878				
2	-0.422214	0.307458	-0.244372				
3	-0.343194	0.307458	-0.184120				
4	1.290626	-1.975729	-0.184120				
...			
1936	-0.228242	1.043971	0.719669				
1937	-0.231353	0.786191	0.237649				
1938	-0.210042	1.080796	0.659417				
1939	-0.101498	0.970319	0.659417				
1940	-0.395621	0.381110	0.237649				
0	Length_of_Conveyer	TypesOfSteel_A300	...	Luminosity_Index	\\		
0	1.454519	1.091255	...	-1.140097			
1	1.454519	1.091255	...	-0.313702			
2	1.028131	1.091255	...	0.063426			
3	-0.770692	-0.915808	...	-0.179421			
4	-0.770692	-0.915808	...	-0.482266			
...			
1936	-0.724056	-0.915808	...	0.959104			
1937	-0.724056	-0.915808	...	0.524836			
1938	-0.724056	-0.915808	...	0.977674			
1939	-0.724056	-0.915808	...	0.818395			
1940	-0.724056	1.091255	...	0.126995			
0	SigmoidOfAreas	Pastry	Z_Scratch	K_Scratch	K_Scratch	Stains	\\
0	0.178899	1	0	0	0	0	
1	-0.703548	1	0	0	0	0	
2	-0.962871	1	0	0	0	0	
3	-0.010774	1	0	0	0	0	
4	1.478005	1	0	0	0	0	
...	
1936	0.624165	0	0	0	0	0	
1937	0.909918	0	0	0	0	0	
1938	0.569750	0	0	0	0	0	
1939	1.452819	0	0	0	0	0	
1940	0.015345	0	0	0	0	0	
0	Dirtiness	Bumps	Class				
0	0	0	1				

```

1          0      0      1
2          0      0      1
3          0      0      1
4          0      0      1
...
...      ...    ...
1936        0      0      2
1937        0      0      2
1938        0      0      2
1939        0      0      2
1940        0      0      2

```

[1613 rows x 31 columns]

```
[15]: df2["Class"]-=1
df2_multiclass = df2.copy()
```

2.1 Feature Selection

2.1.1 Random Forest

```
[16]: from sklearn.ensemble import RandomForestClassifier
from sklearn.feature_selection import SelectFromModel

columns_to_ignore = ["Pastry", "Z_Scratch", "K_Scratch", "Stains", "Dirtiness", ↴
                     "Bumps", "Class"]
df_feature_sel = df2.drop(columns=columns_to_ignore)

# Selecting the target column
target = df2["Class"]

clf = RandomForestClassifier()
clf = clf.fit(df_feature_sel, target)
print(clf.feature_importances_)

model = SelectFromModel(clf, prefit=True)

mask = model.get_support()
selected_columns = df_feature_sel.columns[mask]

df_ranFor = df_feature_sel.loc[:, selected_columns]
```

[0.05613872 0.06180281 0.04461307 0.04047809 0.03535739 0.04931517
 0.05183656 0.04949704 0.06652219 0.01614142 0.01373963 0.07091731
 0.04009099 0.04535289 0.0401377 0.04352764 0.00498507 0.05151231
 0.03446376 0.03216181 0.05537246 0.05887147 0.0371645]

```
[17]: df_feature_sel
```

[17] :	X_Minimum	X_Maximum	Pixels_Areas	X_Perimeter	Y_Perimeter	\
0	-1.135234	-1.194042	-0.273978	-0.488992	0.088914	
1	0.030147	-0.008917	-0.393159	-0.598267	-0.206576	
2	0.385752	0.353917	-0.420893	-0.629489	-0.438747	
3	0.432136	0.403215	-0.342188	-0.551435	0.110021	
4	1.274766	1.282692	1.331600	0.182266	4.647910	
...	
1936	-0.735178	-0.746416	-0.269480	0.088602	-0.375428	
1937	-0.938105	-0.947552	-0.258986	-0.067504	-0.333215	
1938	-0.936172	-0.949524	-0.255238	-0.129947	-0.375428	
1939	-0.951633	-0.957411	-0.160043	0.759861	0.152233	
1940	1.220652	1.233394	-0.396907	-0.348496	-0.375428	
	Sum_of_Luminosity	Minimum_of_Luminosity	Maximum_of_Luminosity			\
0	-0.305749		-0.539531		-1.268667	
1	-0.397664		-0.244926		-0.364878	
2	-0.422214		0.307458		-0.244372	
3	-0.343194		0.307458		-0.184120	
4	1.290626		-1.975729		-0.184120	
...	
1936	-0.228242		1.043971		0.719669	
1937	-0.231353		0.786191		0.237649	
1938	-0.210042		1.080796		0.659417	
1939	-0.101498		0.970319		0.659417	
1940	-0.395621		0.381110		0.237649	
	Length_of_Conveyer	TypesOfSteel_A300	...	Outside_X_Index		\
0	1.454519		1.091255	...	-0.514948	
1	1.454519		1.091255	...	-0.556364	
2	1.028131		1.091255	...	-0.552599	
3	-0.770692		-0.915808	...	-0.496123	
4	-0.770692		-0.915808	...	-0.217504	
...	
1936	-0.724056		-0.915808	...	0.083706	
1937	-0.724056		-0.915808	...	0.166538	
1938	-0.724056		-0.915808	...	0.110061	
1939	-0.724056		-0.915808	...	0.223015	
1940	-0.724056		1.091255	...	-0.138436	
	Edges_X_Index	Edges_Y_Index	Outside_Global_Index	Log0fAreas		\
0	-0.718842	0.695123		0.836722	0.237173	
1	-0.159904	0.526690		0.836722	-0.417695	
2	0.488015	0.429070		0.836722	-0.721056	
3	-0.425551	0.695123		0.836722	-0.064356	
4	-1.527877	0.636955		0.836722	1.828612	
...	
1936	-0.511940	-0.684202		-1.263874	0.253332	

1937	0.291911	-0.780305	-1.263874	0.289482
1938	0.380028	-0.914342	-1.263874	0.301977
1939	-1.282100	-1.026631	-1.263874	0.563191
1940	0.570948	-0.684202	-1.263874	-0.452012
	Log_X_Index	Log_Y_Index	Orientation_Index	Luminosity_Index \
0	-0.846876	0.865673	1.406965	-1.140097
1	-1.190733	0.399819	1.356268	-0.313702
2	-1.190733	-0.132917	1.100962	0.063426
3	-1.006553	0.890625	1.459885	-0.179421
4	0.054474	2.837130	1.640457	-0.482266
...
1936	0.651061	-0.264622	-1.111353	0.959104
1937	0.772746	-0.196712	-1.157808	0.524836
1938	0.692907	-0.336648	-1.220827	0.977674
1939	0.847354	0.474417	-0.368057	0.818395
1940	0.248564	-0.264622	-0.649621	0.126995
	SigmoidOfAreas			
0	0.178899			
1	-0.703548			
2	-0.962871			
3	-0.010774			
4	1.478005			
...	...			
1936	0.624165			
1937	0.909918			
1938	0.569750			
1939	1.452819			
1940	0.015345			

[1613 rows x 23 columns]

```
[18]: df_ranFor["Class"] = df2["Class"]
df_ranFor
```

	X_Minimum	X_Maximum	Pixels_Areas	Sum_of_Luminosity \
0	-1.135234	-1.194042	-0.273978	-0.305749
1	0.030147	-0.008917	-0.393159	-0.397664
2	0.385752	0.353917	-0.420893	-0.422214
3	0.432136	0.403215	-0.342188	-0.343194
4	1.274766	1.282692	1.331600	1.290626
...
1936	-0.735178	-0.746416	-0.269480	-0.228242
1937	-0.938105	-0.947552	-0.258986	-0.231353
1938	-0.936172	-0.949524	-0.255238	-0.210042
1939	-0.951633	-0.957411	-0.160043	-0.101498

1940	1.220652	1.233394	-0.396907	-0.395621									
	Minimum_of_Luminosity	Maximum_of_Luminosity	Length_of_Conveyer	\									
0	-0.539531	-1.268667	1.454519										
1	-0.244926	-0.364878	1.454519										
2	0.307458	-0.244372	1.028131										
3	0.307458	-0.184120	-0.770692										
4	-1.975729	-0.184120	-0.770692										
...										
1936	1.043971	0.719669	-0.724056										
1937	0.786191	0.237649	-0.724056										
1938	1.080796	0.659417	-0.724056										
1939	0.970319	0.659417	-0.724056										
1940	0.381110	0.237649	-0.724056										
	Steel_Plate_Thickness	Outside_X_Index	Edges_Y_Index	LogOfAreas	\								
0	0.035167	-0.514948	0.695123	0.237173									
1	0.035167	-0.556364	0.526690	-0.417695									
2	0.475065	-0.552599	0.429070	-0.721056									
3	4.654098	-0.496123	0.695123	-0.064356									
4	2.344632	-0.217504	0.636955	1.828612									
...								
1936	-0.844629	0.083706	-0.684202	0.253332									
1937	-0.844629	0.166538	-0.780305	0.289482									
1938	-0.844629	0.110061	-0.914342	0.301977									
1939	-0.844629	0.223015	-1.026631	0.563191									
1940	0.035167	-0.138436	-0.684202	-0.452012									
	Orientation_Index	Luminosity_Index	Class										
0	1.406965	-1.140097	0										
1	1.356268	-0.313702	0										
2	1.100962	0.063426	0										
3	1.459885	-0.179421	0										
4	1.640457	-0.482266	0										
...										
1936	-1.111353	0.959104	1										
1937	-1.157808	0.524836	1										
1938	-1.220827	0.977674	1										
1939	-0.368057	0.818395	1										
1940	-0.649621	0.126995	1										

[1613 rows x 14 columns]

2.2 Saving the Data

[19]: df_ranFor

```
[19]:      X_Minimum  X_Maximum  Pixels_Areas  Sum_of_Luminosity \
0      -1.135234  -1.194042   -0.273978    -0.305749
1       0.030147  -0.008917   -0.393159    -0.397664
2       0.385752  0.353917   -0.420893    -0.422214
3       0.432136  0.403215   -0.342188    -0.343194
4       1.274766  1.282692   1.331600     1.290626
...
1936   -0.735178  -0.746416   -0.269480    -0.228242
1937   -0.938105  -0.947552   -0.258986    -0.231353
1938   -0.936172  -0.949524   -0.255238    -0.210042
1939   -0.951633  -0.957411   -0.160043    -0.101498
1940   1.220652   1.233394   -0.396907    -0.395621

      Minimum_of_Luminosity  Maximum_of_Luminosity  Length_of_Conveyer \
0           -0.539531        -1.268667        1.454519
1           -0.244926        -0.364878        1.454519
2            0.307458        -0.244372        1.028131
3            0.307458        -0.184120       -0.770692
4           -1.975729        -0.184120       -0.770692
...
1936          1.043971        0.719669      -0.724056
1937          0.786191        0.237649      -0.724056
1938          1.080796        0.659417      -0.724056
1939          0.970319        0.659417      -0.724056
1940          0.381110        0.237649      -0.724056

      Steel_Plate_Thickness  Outside_X_Index  Edges_Y_Index  LogOfAreas \
0             0.035167        -0.514948        0.695123        0.237173
1             0.035167        -0.556364        0.526690      -0.417695
2             0.475065        -0.552599        0.429070      -0.721056
3             4.654098        -0.496123        0.695123      -0.064356
4             2.344632        -0.217504        0.636955        1.828612
...
1936         -0.844629        0.083706      -0.684202        0.253332
1937         -0.844629        0.166538      -0.780305        0.289482
1938         -0.844629        0.110061      -0.914342        0.301977
1939         -0.844629        0.223015      -1.026631        0.563191
1940          0.035167        -0.138436      -0.684202      -0.452012

      Orientation_Index  Luminosity_Index  Class
0              1.406965      -1.140097        0
1              1.356268      -0.313702        0
2              1.100962       0.063426        0
```

```

3          1.459885      -0.179421      0
4          1.640457      -0.482266      0
...
1936       -1.111353      0.959104      1
1937       -1.157808      0.524836      1
1938       -1.220827      0.977674      1
1939       -0.368057      0.818395      1
1940       -0.649621      0.126995      1

```

[1613 rows x 14 columns]

[20]: df2_multiclass

```

[20]:   X_Minimum  X_Maximum  Pixels_Areas  X_Perimeter  Y_Perimeter \
0     -1.135234  -1.194042  -0.273978  -0.488992  0.088914
1      0.030147  -0.008917  -0.393159  -0.598267  -0.206576
2      0.385752  0.353917  -0.420893  -0.629489  -0.438747
3      0.432136  0.403215  -0.342188  -0.551435  0.110021
4      1.274766  1.282692  1.331600  0.182266  4.647910
...
```

```

1936    -0.735178  -0.746416  -0.269480  0.088602  -0.375428
1937    -0.938105  -0.947552  -0.258986  -0.067504  -0.333215
1938    -0.936172  -0.949524  -0.255238  -0.129947  -0.375428
1939    -0.951633  -0.957411  -0.160043  0.759861  0.152233
1940    1.220652   1.233394  -0.396907  -0.348496  -0.375428

```

```

Sum_of_Luminosity  Minimum_of_Luminosity  Maximum_of_Luminosity \
0                 -0.305749           -0.539531           -1.268667
1                 -0.397664           -0.244926           -0.364878
2                 -0.422214           0.307458           -0.244372
3                 -0.343194           0.307458           -0.184120
4                 1.290626           -1.975729           -0.184120
...
1936       -0.228242           1.043971           0.719669
1937       -0.231353           0.786191           0.237649
1938       -0.210042           1.080796           0.659417
1939       -0.101498           0.970319           0.659417
1940       -0.395621           0.381110           0.237649

```

```

Length_of_Conveyer  TypesOfSteel_A300 ... Luminosity_Index \
0                  1.454519  1.091255 ... -1.140097
1                  1.454519  1.091255 ... -0.313702
2                  1.028131  1.091255 ... 0.063426
3                  -0.770692 -0.915808 ... -0.179421
4                  -0.770692 -0.915808 ... -0.482266
...
1936       -0.724056           -0.915808 ... 0.959104

```

1937	-0.724056	-0.915808	...	0.524836			
1938	-0.724056	-0.915808	...	0.977674			
1939	-0.724056	-0.915808	...	0.818395			
1940	-0.724056	1.091255	...	0.126995			
	SigmoidOfAreas	Pastry	Z_Scratch	K_Scratch	K_Scratch	Stains	\
0	0.178899	1	0	0	0	0	
1	-0.703548	1	0	0	0	0	
2	-0.962871	1	0	0	0	0	
3	-0.010774	1	0	0	0	0	
4	1.478005	1	0	0	0	0	
...	
1936	0.624165	0	0	0	0	0	
1937	0.909918	0	0	0	0	0	
1938	0.569750	0	0	0	0	0	
1939	1.452819	0	0	0	0	0	
1940	0.015345	0	0	0	0	0	
	Dirtiness	Bumps	Class				
0	0	0	0				
1	0	0	0				
2	0	0	0				
3	0	0	0				
4	0	0	0				
...				
1936	0	0	1				
1937	0	0	1				
1938	0	0	1				
1939	0	0	1				
1940	0	0	1				

[1613 rows x 31 columns]

```
[21]: df_ranFor.to_csv("data/norm_data.csv", index=False)
df2_multiclass.to_csv("data/norm_multiclass_data.csv", index=False)
```