```
In [1]:
                                                                                     H
 1 import pandas as pd
 2 import numpy as np
 3 import matplotlib.pyplot as plt
 4 import seaborn as sns
 5 from IPython import get_ipython
 6 import warnings
 7 warnings.filterwarnings("ignore")
In [2]:
```

1 | fetal_data = pd.read_csv('fetal_health.csv')

In [3]: 1 fetal_data.head()

Out[3]:

	baseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	severe_dece
0	120	0.000	0.0	0.000	0.000	_
1	132	0.006	0.0	0.006	0.003	
2	133	0.003	0.0	0.008	0.003	
3	134	0.003	0.0	0.008	0.003	
4	132	0.007	0.0	0.008	0.000	

5 rows × 22 columns

In [4]: 1 fetal_data.tail()

Out[4]:

	baseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	severe_c
2121	140	0.000	0.000	0.007	0.0	
2122	140	0.001	0.000	0.007	0.0	
2123	140	0.001	0.000	0.007	0.0	
2124	140	0.001	0.000	0.006	0.0	
2125	142	0.002	0.002	0.008	0.0	

5 rows × 22 columns

H

```
In [5]:
   fetal_data.shape
Out[5]:
(2126, 22)
In [6]:
                                                                                        M
   fetal_data.columns
Out[6]:
Index(['baseline value', 'accelerations', 'fetal_movement',
       'uterine_contractions', 'light_decelerations', 'severe_deceleration
s',
       'prolongued_decelerations', 'abnormal_short_term_variability',
       'mean_value_of_short_term_variability',
       'percentage_of_time_with_abnormal_long_term_variability',
       'mean_value_of_long_term_variability', 'histogram_width',
       'histogram_min', 'histogram_max', 'histogram_number_of_peaks',
       'histogram_number_of_zeroes', 'histogram_mode', 'histogram_mean',
       'histogram_median', 'histogram_variance', 'histogram_tendency',
       'fetal health'],
      dtype='object')
In [7]:
                                                                                        H
   fetal_data.duplicated().sum()
Out[7]:
13
                                                                                        M
In [8]:
   fetal_data = fetal_data.drop_duplicates()
In [9]:
                                                                                        M
   fetal data.shape
Out[9]:
(2113, 22)
```

In [10]:

```
1 fetal_data.isnull().sum()
```

Out[10]:

baseline value	0
accelerations	0
fetal_movement	0
uterine_contractions	0
light_decelerations	0
severe_decelerations	0
<pre>prolongued_decelerations</pre>	0
abnormal_short_term_variability	0
<pre>mean_value_of_short_term_variability</pre>	0
<pre>percentage_of_time_with_abnormal_long_term_variability</pre>	0
mean_value_of_long_term_variability	0
histogram_width	0
histogram_min	0
histogram_max	0
histogram_number_of_peaks	0
histogram_number_of_zeroes	0
histogram_mode	0
histogram_mean	0
histogram_median	0
histogram_variance	0
histogram_tendency	0
fetal_health	0
dtype: int64	

In [11]:

```
1 fetal_data.info()
```

<pre><class 'pandas.core.frame.dataframe'=""> Int64Index: 2113 entries, 0 to 2125 Data columns (total 22 columns): # Column</class></pre>	Non-Null Coun
t Dtype	Non Nail Coun
0 baseline value	2113 non-null
int64	
1 accelerations	2113 non-null
float64	
2 fetal_movement	2113 non-null
float64	
<pre>3 uterine_contractions</pre>	2113 non-null
float64	
4 light_decelerations	2113 non-null
float64	2442
5 severe_decelerations	2113 non-null
float64 6 prolongued decelerations	2113 non-null
float64	2113 HOH-HUII
7 abnormal_short_term_variability	2113 non-null
int64	ZIIJ NON-NUII
8 mean_value_of_short_term_variability	2113 non-null
float64	2223 11011 11022
<pre>9 percentage_of_time_with_abnormal_long_term_variability</pre>	2113 non-null
int64	
<pre>10 mean_value_of_long_term_variability</pre>	2113 non-null
float64	
11 histogram_width	2113 non-null
int64	
12 histogram_min	2113 non-null
int64	
13 histogram_max	2113 non-null
int64	
14 histogram_number_of_peaks	2113 non-null
int64	211211
<pre>15 histogram_number_of_zeroes int64</pre>	2113 non-null
16 histogram mode	2113 non-null
int64	ZII3 HOH-HUII
17 histogram_mean	2113 non-null
int64	ZIIJ NON NUII
18 histogram median	2113 non-null
int64	
19 histogram_variance	2113 non-null
int64	
20 histogram_tendency	2113 non-null
int64	
21 fetal_health	2113 non-null
int64	
dtypes: float64(8), int64(14)	
memory usage: 379.7 KB	

In [12]:

1 fetal_data.describe()

Out[12]:

	baseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	sevi
count	2113.000000	2113.000000	2113.000000	2113.000000	2113.000000	
mean	133.304780	0.003188	0.009517	0.004387	0.001901	
std	9.837451	0.003871	0.046804	0.002941	0.002966	
min	106.000000	0.000000	0.000000	0.000000	0.000000	
25%	126.000000	0.000000	0.000000	0.002000	0.000000	
50%	133.000000	0.002000	0.000000	0.005000	0.000000	
75%	140.000000	0.006000	0.003000	0.007000	0.003000	
max	160.000000	0.019000	0.481000	0.015000	0.015000	
8 rows × 22 columns						
4						•

In [13]:

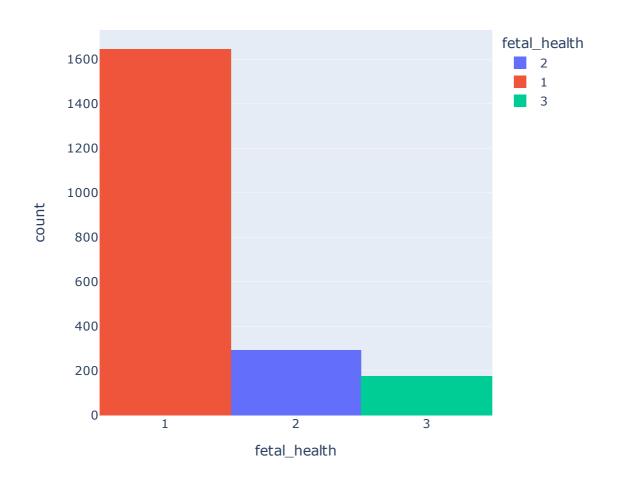
1 fetal_data.nunique()

Out[13]:

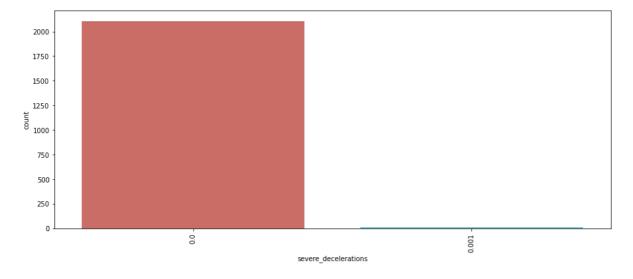
baseline value	48
accelerations	20
fetal_movement	102
uterine_contractions	16
light_decelerations	16
severe decelerations	2
prolongued decelerations	6
abnormal_short_term_variability	75
mean_value_of_short_term_variability	57
<pre>percentage_of_time_with_abnormal_long_term_variability</pre>	87
mean_value_of_long_term_variability	249
histogram_width	154
histogram_min	109
histogram_max	86
histogram_number_of_peaks	18
histogram_number_of_zeroes	9
histogram mode	88
histogram mean	103
histogram_median	95
histogram_variance	133
-	
histogram_tendency	3
fetal_health	3
dtype: int64	

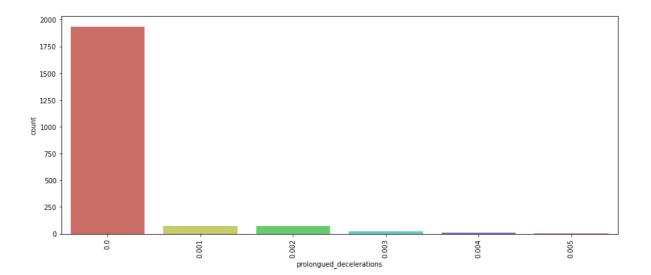
```
H
In [14]:
 1 fetal_data['fetal_health'].unique()
Out[14]:
array([2, 1, 3], dtype=int64)
In [15]:
                                                                                        H
 1 fetal_data['fetal_health'].value_counts()
Out[15]:
1
     1646
      292
2
3
      175
Name: fetal_health, dtype: int64
In [16]:
   import plotly.express as px
```

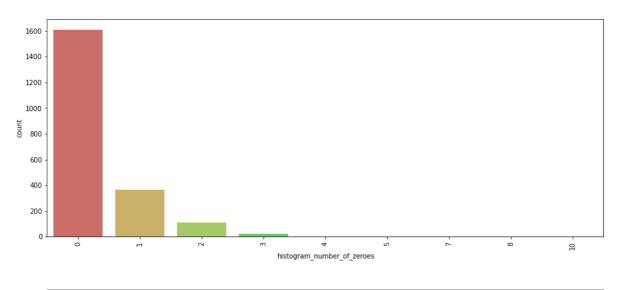
In [17]: ▶

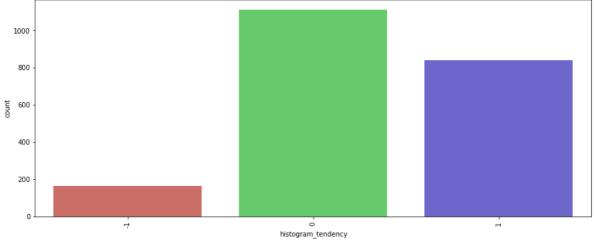


In [19]:



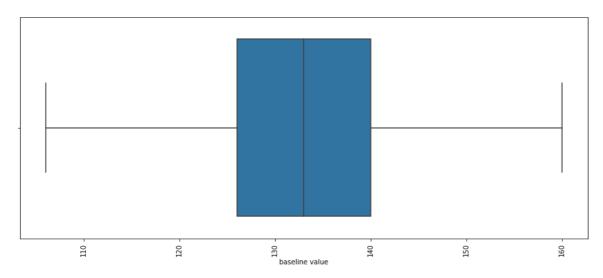






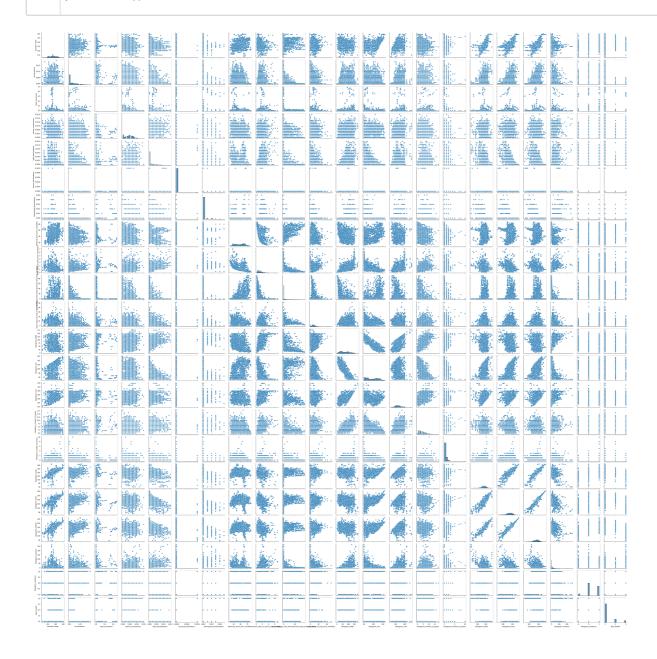
In [20]: ▶

```
for i in fetal_data.columns:
   plt.figure(figsize=(15,6))
   sns.boxplot(fetal_data[i], data = fetal_data)
   plt.xticks(rotation = 90)
   plt.show()
```



H In [21]:

- sns.pairplot(fetal_data) 1 2
 - plt.show()



In [23]: ▶

1 fetal_data.corr()

Out[23]:

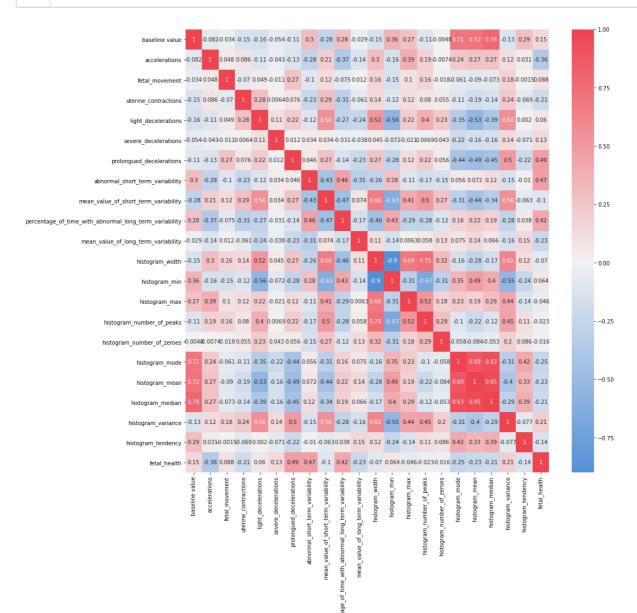
	baseline value	accelerations	fetal_movement
baseline value	1.000000	-0.081885	-0.033949
accelerations	-0.081885	1.000000	0.048114
fetal_movement	-0.033949	0.048114	1.000000
uterine_contractions	-0.149587	0.086174	-0.069867
light_decelerations	-0.159836	-0.110595	0.048795
severe_decelerations	-0.053706	-0.043237	-0.011022
prolongued_decelerations	-0.105003	-0.128682	0.265802
abnormal_short_term_variability	0.303502	-0.280495	-0.104876
mean_value_of_short_term_variability	-0.278344	0.206762	0.121532
percentage_of_time_with_abnormal_long_term_variability	0.283918	-0.373507	-0.074900
mean_value_of_long_term_variability	-0.028901	-0.141413	0.011749
histogram_width	-0.147150	0.298350	0.162803
histogram_min	0.360129	-0.155306	-0.154297
histogram_max	0.273402	0.392684	0.099703
histogram_number_of_peaks	-0.113242	0.189209	0.164645
histogram_number_of_zeroes	-0.004807	-0.007360	-0.018122
histogram_mode	0.708074	0.243083	-0.061496
histogram_mean	0.722152	0.270266	-0.089938
histogram_median	0.788487	0.272507	-0.072676
histogram_variance	-0.134458	0.124433	0.179115
histogram_tendency	0.294412	0.030670	-0.001459
fetal_health	0.146077	-0.363947	0.088057

22 rows × 22 columns

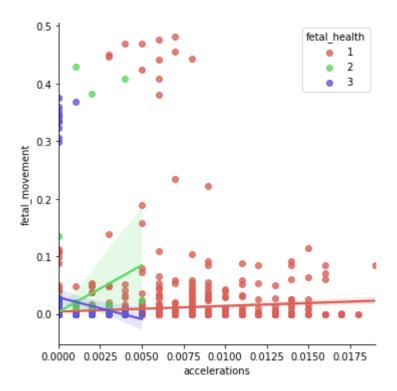
localhost:8888/notebooks/Fetal Health Classification.ipynb

In [25]: ▶

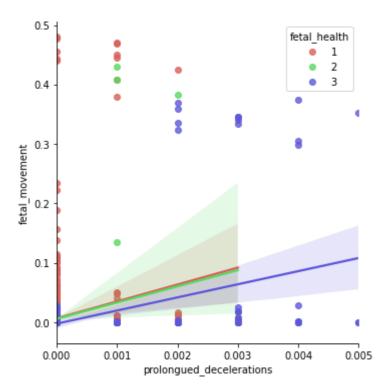
```
plt.figure(figsize=(15,15))
cmap = sns.diverging_palette(250, 10, s=80, l=55, n=9, as_cmap=True)
sns.heatmap(fetal_data.corr(),annot=True, cmap=cmap, center=0)
plt.show()
```



In [27]: ▶

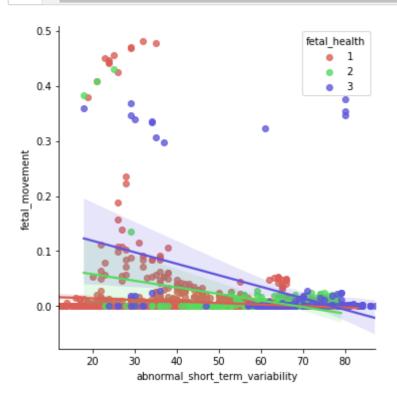


In [29]: ▶



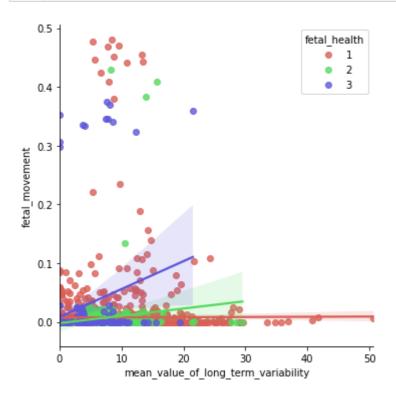
```
In [30]:
```

```
sns.lmplot(data = fetal_data,x="abnormal_short_term_variability",y="fetal_movement"
palette='hls', hue="fetal_health",legend_out=False)
plt.show()
```



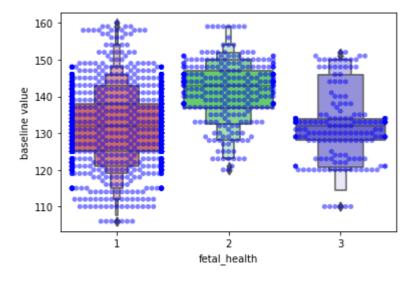
In [31]:

```
sns.lmplot(data = fetal_data,x="mean_value_of_long_term_variability",y="fetal_movem
palette='hls', hue="fetal_health",legend_out=False)
plt.show()
```



```
In [32]: ▶
```

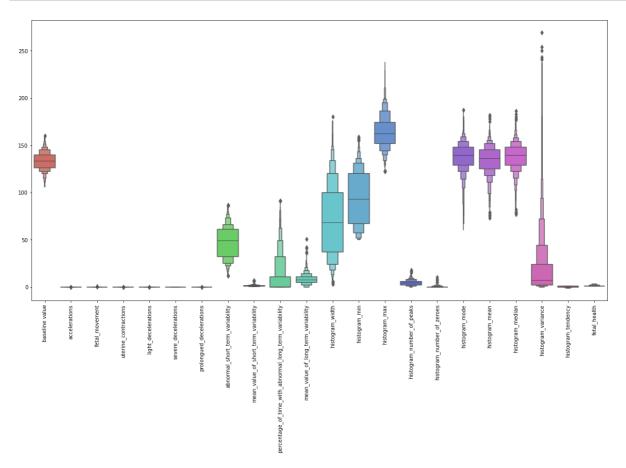
In [33]:





In [34]:

```
plt.figure(figsize=(20,10))
sns.boxenplot(data = fetal_data,palette = 'hls')
plt.xticks(rotation=90)
plt.show()
```



```
In [35]: ▶
```

```
from sklearn.model_selection import train_test_split
   from sklearn import preprocessing
   from sklearn.preprocessing import StandardScaler
   from sklearn.pipeline import Pipeline
 5
   from sklearn.linear_model import LogisticRegression
   from sklearn.tree import DecisionTreeClassifier
 7
   from sklearn.ensemble import RandomForestClassifier
   from sklearn.svm import SVC
 9
   from sklearn.svm import LinearSVC
10 from sklearn.model_selection import GridSearchCV
   from sklearn.model_selection import cross_val_score
   from sklearn.metrics import precision_score, recall_score, confusion_matrix, classi
12
   from sklearn import metrics
   from sklearn.metrics import roc_curve, auc, roc_auc_score
14
   np.random.seed(0)
15
```

```
In [36]:
```

```
1 X=fetal_data.drop(["fetal_health"],axis=1)
2 y=fetal_data["fetal_health"]
```

In	[37]:	H
1	X.shape	
Out	[37]:	
(21	13, 21)	
In	[38]:	K
	[38]: y.shape	H
1		K

In [39]: ▶

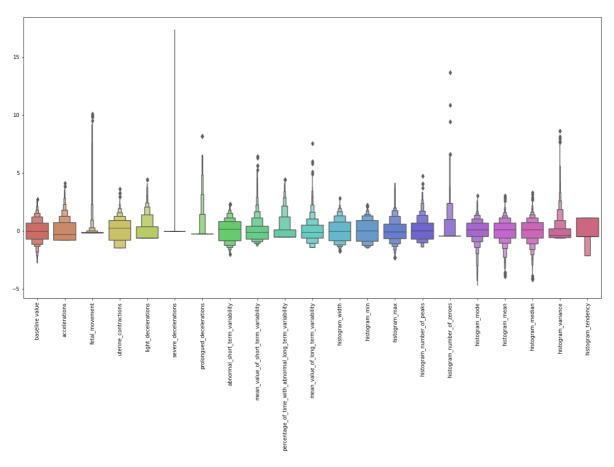
```
col_names = list(X.columns)
s_scaler = preprocessing.StandardScaler()
X_df= s_scaler.fit_transform(X)
X_df = pd.DataFrame(X_df, columns=col_names)
X_df.describe().T
```

Out[39]:

	count	mean	std	m
baseline value	2113.0	-1.143325e- 15	1.000237	-2.77625
accelerations	2113.0	6.725440e-18	1.000237	-0.82377
fetal_movement	2113.0	-1.345088e- 17	1.000237	-0.2033§
uterine_contractions	2113.0	8.070528e-17	1.000237	-1.49224
light_decelerations	2113.0	-8.070528e- 17	1.000237	-0.6412(
severe_decelerations	2113.0	-6.725440e- 18	1.000237	-0.05765
prolongued_decelerations	2113.0	4.035264e-17	1.000237	-0.26964
abnormal_short_term_variability	2113.0	-1.345088e- 17	1.000237	-2.03764
mean_value_of_short_term_variability	2113.0	-4.035264e- 17	1.000237	-1.28373
percentage_of_time_with_abnormal_long_term_variability	2113.0	2.690176e-17	1.000237	-0.53429
mean_value_of_long_term_variability	2113.0	1.076070e-16	1.000237	-1.45015
histogram_width	2113.0	5.044080e-17	1.000237	-1.73174
histogram_min	2113.0	1.042443e-16	1.000237	-1.47400
histogram_max	2113.0	3.127330e-16	1.000237	-2.34658
histogram_number_of_peaks	2113.0	0.000000e+00	1.000237	-1.38162
histogram_number_of_zeroes	2113.0	7.313916e-17	1.000237	-0.46015
histogram_mode	2113.0	-1.614106e- 16	1.000237	-4.7233€
histogram_mean	2113.0	3.497229e-16	1.000237	-3.9469§
histogram_median	2113.0	3.497229e-16	1.000237	-4.22018
histogram_variance	2113.0	-6.725440e- 17	1.000237	-0.6512{
histogram_tendency	2113.0	2.690176e-17	1.000237	-2.15819
4				>

In [40]:

```
plt.figure(figsize=(20,10))
sns.boxenplot(data = X_df,palette = 'hls')
plt.xticks(rotation=90)
plt.show()
```



```
In [41]:
```

```
In [42]: ▶
```

```
In [43]:
                                                                                        M
    for pipe in pipelines:
 1
        pipe.fit(X_train, y_train)
 2
In [44]:
    cv_results_accuracy = []
 2
    for i, model in enumerate(pipelines):
        cv_score = cross_val_score(model, X_train,y_train, cv=10 )
 3
        cv_results_accuracy.append(cv_score)
 4
 5
        print("%s: %f " % (pipe_dict[i], cv_score.mean()))
Logistic Regression: 0.887574
Decision Tree: 0.915385
RandomForest: 0.937278
SVC: 0.904734
In [45]:
                                                                                        M
 1 pred_rfc = pipeline_rf.predict(X_test)
In [46]:
 1 print("Training Accuracy :", pipeline_rf.score(X_train, y_train))
   print("Testing Accuracy :", pipeline_rf.score(X_test, y_test))
Training Accuracy: 0.9994082840236687
Testing Accuracy : 0.966903073286052
                                                                                        H
In [50]:
   from sklearn.metrics import accuracy_score
In [51]:
                                                                                        Ы
report for classifier %s:\n%s\n" % (pipeline_rf,
               2
                                      metrics.classification report(y test, pred rfc)))
 Classification report for classifier Pipeline(steps=[('rf_classifier', Ra
ndomForestClassifier())]):
              precision
                           recall f1-score
                                               support
                   0.97
                             0.99
                                        0.98
           1
                                                   341
                   0.93
                              0.80
           2
                                        0.86
                                                    54
           3
                   1.00
                                                    28
                              1.00
                                        1.00
    accuracy
                                        0.97
                                                   423
                   0.97
                             0.93
                                        0.95
                                                   423
   macro avg
weighted avg
                   0.97
                             0.97
                                        0.97
                                                   423
```

In [52]: ▶

```
matrix = confusion_matrix(y_test, pred_rfc)
print('Confusion matrix : \n',matrix)
```

In [53]: ▶

```
plt.subplots(figsize=(12,8))
sns.heatmap(matrix/np.sum(matrix), cmap=cmap,annot = True,
annot_kws = {'size':15})
plt.show()
```

