

Based on the program developed in the lecture, implement a nonlinear Kernel analysis based on SVM for (a) polynomial kernels (poly) of at least 2 different degrees of your choice, and (b) for radial base functions (rbf). Apply the analysis to the breast cancer database, or a database of your choice. Study the classification performance in terms of suitable measures of your choice (e.g., accuracy, precision, recall, f1, k-folds) as a function of the hyperparameters γ (gamma) and degree (for polynomials), and gamma and penalty C for rbf kernels. It is on you if you want to use GridSearch, RandomSearch, hyperopt, or just to try a few combinations of the hyperparameters. Hint: Hyperopt may require you to install or update some libraries outside the Jupyter notebook. The program must end with the output "Hyper, hyper!".

0. Imports, global variables/constants, and datasets

```
In [1]: from sklearn import svm

import pandas as pd

from sklearn.metrics import precision_recall_fscore_support

import numpy as np

# Load datasets
from sklearn import datasets
cancer = datasets.load_breast_cancer()
print("Features: ", cancer.feature_names)
print("Labels: ", cancer.target_names)
print(cancer.data.shape)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(
    cancer.data,
    cancer.target,
    test_size=0.3,
    random_state=77
)

import warnings
warnings.filterwarnings("ignore")
```

```
Features: ['mean radius' 'mean texture' 'mean perimeter' 'mean area'
'mean smoothness' 'mean compactness' 'mean concavity'
'mean concave points' 'mean symmetry' 'mean fractal dimension'
'radius error' 'texture error' 'perimeter error' 'area error'
'smoothness error' 'compactness error' 'concavity error'
'concave points error' 'symmetry error' 'fractal dimension error'
'worst radius' 'worst texture' 'worst perimeter' 'worst area'
'worst smoothness' 'worst compactness' 'worst concavity'
'worst concave points' 'worst symmetry' 'worst fractal dimension']
Labels: ['malignant' 'benign']
(569, 30)
```

1. (a) Nonlinear Kernel analysis based on SVM for polynomial kernels, 2 degrees

```
In [2]: clf_poly = svm.SVC(  
        kernel='poly',  
        degree=2,  
        gamma='auto'  
    )
```

2. (b) Nonlinear Kernel analysis based on SVM for radial base functions

```
In [3]: clf_rbf = svm.SVC(kernel='rbf')
```

3. Apply analysis to the breast cancer database (or another database)

```
In [4]: clf_poly.fit(X_train, y_train)  
        clf_rbf.fit(X_train, y_train)
```

```
Out[4]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,  
            decision_function_shape='ovr', degree=3, gamma='auto_deprecated',  
            kernel='rbf', max_iter=-1, probability=False, random_state=None,  
            shrinking=True, tol=0.001, verbose=False)
```

4. Study classification performance (accuracy, precision, recall, f1, k-folds) as a function of the hyperparameters γ and degree (for polynomials), and γ and penalty C for rbf kernels.

I decided to try different parameters and plot the test scores.

```

In [5]: def get_accuracy_precision_recall_f_score(clf_instance, X_train, y_train, X_test, y_test):
    clf_instance.fit(
        X_train,
        y_train
    )

    y_pred = clf_instance.predict(X_test)

    accuracy = clf_instance.score(X_test, y_test)

    precision, recall, f_score, _ = precision_recall_fscore_support(y_test, y_pred)

    return accuracy, precision, recall, f_score

def analyze_svm(hyperparameters, X_train, y_train, X_test, y_test, polynomials=True):
    df_results = pd.DataFrame.from_dict(
        {
            'accuracy': [],
            'precision': [],
            'recall': [],
            'f_score': [],
            'gamma': [],
            'degree': [],
            'penalty_c': []
        }
    )

    if polynomials:
        for gamma in hyperparameters['gamma']:
            for degree in hyperparameters['degree']:

                clf_poly = svm.SVC(
                    kernel='poly',
                    degree=degree,
                    gamma=gamma
                )

                accuracy, precision, recall, f_score = get_accuracy_precision_recall_f_score(
                    clf_poly,
                    X_train,
                    y_train,
                    X_test,
                    y_test
                )

                df_intermediate = pd.DataFrame.from_dict(
                    {
                        'accuracy': accuracy,
                        'precision': precision,
                        'recall': precision,
                        'f_score': f_score,
                        'gamma': gamma,

```

```

        'degree': degree,
        'penalty_c': None
    }
)
df_results = df_results.append(
    df_intermediate,
    ignore_index=True
)

else:
    for gamma in hyperparameters['gamma']:
        for penalty_c in hyperparameters['penalty_c']:

            clf_rbf = svm.SVC(
                kernel='rbf',
                gamma=gamma,
                C=penalty_c
            )

            accuracy, precision, recall, f_score = get_accuracy_p
recision_recall_f_score(
                clf_rbf,
                X_train,
                y_train,
                X_test,
                y_test
            )

            df_intermediate = pd.DataFrame.from_dict(
                {
                    'accuracy': accuracy,
                    'precision': precision,
                    'recall': precision,
                    'f_score': f_score,
                    'gamma': gamma,
                    'degree': None,
                    'penalty_c': penalty_c
                }
            )
            df_results = df_results.append(
                df_intermediate,
                ignore_index=True
            )

    return df_results

list_gamma = ['scale', 'auto', 'auto_deprecated']
#list_gamma = [0.1, 1]#, 'auto', 'auto_deprecated']

hyperparameters = {
    'gamma': list_gamma,
    'degree': [0, 1, 2]#, 3]
}
df_results = analyze_svm(hyperparameters, X_train, y_train, X_test, y
_test, polynomials=True)
print('***POLY***')
print('Sorted by accuracy')

```

```
print(df_results.sort_values(by='accuracy', ascending=False))
print('\nSorted by precision')
print(df_results.sort_values(by='precision', ascending=False))
print('\nSorted by recall')
print(df_results.sort_values(by='recall', ascending=False))
print('\nSorted by f-score')
print(df_results.sort_values(by='f_score', ascending=False))

hyperparameters = {
    'gamma': list_gamma,
    'penalty_c': [0.1, 1, 10]#, 100]
}
df_results = analyze_svm(hyperparameters, X_train, y_train, X_test, y_test, polynomials=False)
print('\n\n\n')
print('***RBF***')
print('Sorted by accuracy')
print(df_results.sort_values(by='accuracy', ascending=False))
print('\nSorted by precision')
print(df_results.sort_values(by='precision', ascending=False))
print('\nSorted by recall')
print(df_results.sort_values(by='recall', ascending=False))
print('\nSorted by f-score')
print(df_results.sort_values(by='f_score', ascending=False))
```

POLY

Sorted by accuracy

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
17	0.970760	0.964286	0.964286	0.977376	auto_deprecated	2.0
	None					
16	0.970760	0.983051	0.983051	0.958678	auto_deprecated	2.0
	None					
11	0.970760	0.964286	0.964286	0.977376	auto	2.0
	None					
10	0.970760	0.983051	0.983051	0.958678	auto	2.0
	None					
8	0.964912	0.966667	0.966667	0.950820	auto	1.0
	None					
15	0.964912	0.963964	0.963964	0.972727	auto_deprecated	1.0
	None					
14	0.964912	0.966667	0.966667	0.950820	auto_deprecated	1.0
	None					
9	0.964912	0.963964	0.963964	0.972727	auto	1.0
	None					
5	0.906433	0.878049	0.878049	0.931034	scale	2.0
	None					
4	0.906433	0.979167	0.979167	0.854545	scale	2.0
	None					
3	0.900585	0.870968	0.870968	0.927039	scale	1.0
	None					
2	0.900585	0.978723	0.978723	0.844037	scale	1.0
	None					
7	0.637427	0.637427	0.637427	0.778571	auto	0.0
	None					
1	0.637427	0.637427	0.637427	0.778571	scale	0.0
	None					
6	0.637427	0.000000	0.000000	0.000000	auto	0.0
	None					
12	0.637427	0.000000	0.000000	0.000000	auto_deprecated	0.0
	None					
13	0.637427	0.637427	0.637427	0.778571	auto_deprecated	0.0
	None					
0	0.637427	0.000000	0.000000	0.000000	scale	0.0
	None					

Sorted by precision

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
10	0.970760	0.983051	0.983051	0.958678	auto	2.0
	None					
16	0.970760	0.983051	0.983051	0.958678	auto_deprecated	2.0
	None					
4	0.906433	0.979167	0.979167	0.854545	scale	2.0
	None					
2	0.900585	0.978723	0.978723	0.844037	scale	1.0
	None					
14	0.964912	0.966667	0.966667	0.950820	auto_deprecated	1.0
	None					
8	0.964912	0.966667	0.966667	0.950820	auto	1.0
	None					
17	0.970760	0.964286	0.964286	0.977376	auto_deprecated	2.0

	None					
11	0.970760	0.964286	0.964286	0.977376	auto	2.0
	None					
15	0.964912	0.963964	0.963964	0.972727	auto_deprecated	1.0
	None					
9	0.964912	0.963964	0.963964	0.972727	auto	1.0
	None					
5	0.906433	0.878049	0.878049	0.931034	scale	2.0
	None					
3	0.900585	0.870968	0.870968	0.927039	scale	1.0
	None					
1	0.637427	0.637427	0.637427	0.778571	scale	0.0
	None					
7	0.637427	0.637427	0.637427	0.778571	auto	0.0
	None					
13	0.637427	0.637427	0.637427	0.778571	auto_deprecated	0.0
	None					
12	0.637427	0.000000	0.000000	0.000000	auto_deprecated	0.0
	None					
6	0.637427	0.000000	0.000000	0.000000	auto	0.0
	None					
0	0.637427	0.000000	0.000000	0.000000	scale	0.0
	None					

Sorted by recall

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
10	0.970760	0.983051	0.983051	0.958678	auto	2.0
	None					
16	0.970760	0.983051	0.983051	0.958678	auto_deprecated	2.0
	None					
4	0.906433	0.979167	0.979167	0.854545	scale	2.0
	None					
2	0.900585	0.978723	0.978723	0.844037	scale	1.0
	None					
14	0.964912	0.966667	0.966667	0.950820	auto_deprecated	1.0
	None					
8	0.964912	0.966667	0.966667	0.950820	auto	1.0
	None					
17	0.970760	0.964286	0.964286	0.977376	auto_deprecated	2.0
	None					
11	0.970760	0.964286	0.964286	0.977376	auto	2.0
	None					
15	0.964912	0.963964	0.963964	0.972727	auto_deprecated	1.0
	None					
9	0.964912	0.963964	0.963964	0.972727	auto	1.0
	None					
5	0.906433	0.878049	0.878049	0.931034	scale	2.0
	None					
3	0.900585	0.870968	0.870968	0.927039	scale	1.0
	None					
1	0.637427	0.637427	0.637427	0.778571	scale	0.0
	None					
7	0.637427	0.637427	0.637427	0.778571	auto	0.0
	None					
13	0.637427	0.637427	0.637427	0.778571	auto_deprecated	0.0
	None					

12	0.637427 None	0.000000	0.000000	0.000000	auto_deprecated	0.0
6	0.637427 None	0.000000	0.000000	0.000000	auto	0.0
0	0.637427 None	0.000000	0.000000	0.000000	scale	0.0

Sorted by f-score

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
17	0.970760 None	0.964286	0.964286	0.977376	auto_deprecated	2.0
11	0.970760 None	0.964286	0.964286	0.977376	auto	2.0
15	0.964912 None	0.963964	0.963964	0.972727	auto_deprecated	1.0
9	0.964912 None	0.963964	0.963964	0.972727	auto	1.0
10	0.970760 None	0.983051	0.983051	0.958678	auto	2.0
16	0.970760 None	0.983051	0.983051	0.958678	auto_deprecated	2.0
8	0.964912 None	0.966667	0.966667	0.950820	auto	1.0
14	0.964912 None	0.966667	0.966667	0.950820	auto_deprecated	1.0
5	0.906433 None	0.878049	0.878049	0.931034	scale	2.0
3	0.900585 None	0.870968	0.870968	0.927039	scale	1.0
4	0.906433 None	0.979167	0.979167	0.854545	scale	2.0
2	0.900585 None	0.978723	0.978723	0.844037	scale	1.0
7	0.637427 None	0.637427	0.637427	0.778571	auto	0.0
1	0.637427 None	0.637427	0.637427	0.778571	scale	0.0
13	0.637427 None	0.637427	0.637427	0.778571	auto_deprecated	0.0
6	0.637427 None	0.000000	0.000000	0.000000	auto	0.0
12	0.637427 None	0.000000	0.000000	0.000000	auto_deprecated	0.0
0	0.637427 None	0.000000	0.000000	0.000000	scale	0.0

RBF

Sorted by accuracy

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
4	0.941520 10.0	0.964286	0.964286	0.915254	scale	None
5	0.941520	0.930435	0.930435	0.955357	scale	None

	10.0					
2	0.912281	0.979592	0.979592	0.864865	scale	None
	1.0					
3	0.912281	0.885246	0.885246	0.935065	scale	None
	1.0					
0	0.894737	0.978261	0.978261	0.833333	scale	None
	0.1					
1	0.894737	0.864000	0.864000	0.923077	scale	None
	0.1					
12	0.637427	0.000000	0.000000	0.000000	auto_deprecated	None
	0.1					
16	0.637427	0.000000	0.000000	0.000000	auto_deprecated	None
	10.0					
15	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
	1.0					
14	0.637427	0.000000	0.000000	0.000000	auto_deprecated	None
	1.0					
13	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
	0.1					
9	0.637427	0.637427	0.637427	0.778571	auto	None
	1.0					
11	0.637427	0.637427	0.637427	0.778571	auto	None
	10.0					
10	0.637427	0.000000	0.000000	0.000000	auto	None
	10.0					
8	0.637427	0.000000	0.000000	0.000000	auto	None
	1.0					
7	0.637427	0.637427	0.637427	0.778571	auto	None
	0.1					
6	0.637427	0.000000	0.000000	0.000000	auto	None
	0.1					
17	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
	10.0					

Sorted by precision

	accuracy	precision	recall	f_score	gamma	degree
	penalty_c					
2	0.912281	0.979592	0.979592	0.864865	scale	None
	1.0					
0	0.894737	0.978261	0.978261	0.833333	scale	None
	0.1					
4	0.941520	0.964286	0.964286	0.915254	scale	None
	10.0					
5	0.941520	0.930435	0.930435	0.955357	scale	None
	10.0					
3	0.912281	0.885246	0.885246	0.935065	scale	None
	1.0					
1	0.894737	0.864000	0.864000	0.923077	scale	None
	0.1					
15	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
	1.0					
13	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
	0.1					
11	0.637427	0.637427	0.637427	0.778571	auto	None
	10.0					
9	0.637427	0.637427	0.637427	0.778571	auto	None
	1.0					

7	0.637427 0.1	0.637427	0.637427	0.778571	auto	None
17	0.637427 10.0	0.637427	0.637427	0.778571	auto_deprecated	None
10	0.637427 10.0	0.000000	0.000000	0.000000	auto	None
8	0.637427 1.0	0.000000	0.000000	0.000000	auto	None
12	0.637427 0.1	0.000000	0.000000	0.000000	auto_deprecated	None
14	0.637427 1.0	0.000000	0.000000	0.000000	auto_deprecated	None
6	0.637427 0.1	0.000000	0.000000	0.000000	auto	None
16	0.637427 10.0	0.000000	0.000000	0.000000	auto_deprecated	None

Sorted by recall

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
2	0.912281 1.0	0.979592	0.979592	0.864865	scale	None
0	0.894737 0.1	0.978261	0.978261	0.833333	scale	None
4	0.941520 10.0	0.964286	0.964286	0.915254	scale	None
5	0.941520 10.0	0.930435	0.930435	0.955357	scale	None
3	0.912281 1.0	0.885246	0.885246	0.935065	scale	None
1	0.894737 0.1	0.864000	0.864000	0.923077	scale	None
15	0.637427 1.0	0.637427	0.637427	0.778571	auto_deprecated	None
13	0.637427 0.1	0.637427	0.637427	0.778571	auto_deprecated	None
11	0.637427 10.0	0.637427	0.637427	0.778571	auto	None
9	0.637427 1.0	0.637427	0.637427	0.778571	auto	None
7	0.637427 0.1	0.637427	0.637427	0.778571	auto	None
17	0.637427 10.0	0.637427	0.637427	0.778571	auto_deprecated	None
10	0.637427 10.0	0.000000	0.000000	0.000000	auto	None
8	0.637427 1.0	0.000000	0.000000	0.000000	auto	None
12	0.637427 0.1	0.000000	0.000000	0.000000	auto_deprecated	None
14	0.637427 1.0	0.000000	0.000000	0.000000	auto_deprecated	None
6	0.637427 0.1	0.000000	0.000000	0.000000	auto	None
16	0.637427 10.0	0.000000	0.000000	0.000000	auto_deprecated	None

Sorted by f-score

	accuracy	precision	recall	f_score	gamma	degree
penalty_c						
5	0.941520	0.930435	0.930435	0.955357	scale	None
10.0						
3	0.912281	0.885246	0.885246	0.935065	scale	None
1.0						
1	0.894737	0.864000	0.864000	0.923077	scale	None
0.1						
4	0.941520	0.964286	0.964286	0.915254	scale	None
10.0						
2	0.912281	0.979592	0.979592	0.864865	scale	None
1.0						
0	0.894737	0.978261	0.978261	0.833333	scale	None
0.1						
15	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
1.0						
13	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
0.1						
11	0.637427	0.637427	0.637427	0.778571	auto	None
10.0						
9	0.637427	0.637427	0.637427	0.778571	auto	None
1.0						
7	0.637427	0.637427	0.637427	0.778571	auto	None
0.1						
17	0.637427	0.637427	0.637427	0.778571	auto_deprecated	None
10.0						
10	0.637427	0.000000	0.000000	0.000000	auto	None
10.0						
8	0.637427	0.000000	0.000000	0.000000	auto	None
1.0						
12	0.637427	0.000000	0.000000	0.000000	auto_deprecated	None
0.1						
14	0.637427	0.000000	0.000000	0.000000	auto_deprecated	None
1.0						
6	0.637427	0.000000	0.000000	0.000000	auto	None
0.1						
16	0.637427	0.000000	0.000000	0.000000	auto_deprecated	None
10.0						

5. Hyper, hyper!

```
In [6]: print('Hyper, hyper!')
```

```
Hyper, hyper!
```