

In [2]:

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.svm import SVR
from sklearn.metrics import mean_squared_error
from sklearn.model_selection import RandomizedSearchCV, GridSearchCV
```

Lendo arquivos

In [3]:

```
x_train = np.load('Xtreino5.npy')
y_train = np.load('ytreino5.npy')

x_test = np.load('Xteste5.npy')
y_test = np.load('yteste5.npy')

#função para reportar os resultados das combinações de parametros
def report(results, n_top=3):
    for i in range(1, n_top + 1):
        candidates = np.flatnonzero(results['rank_test_score'] == i)
        for candidate in candidates:
            print("Model with rank: {0}".format(i))
            print("Mean validation score: {0:.3f} (std: {1:.3f})".format(
                results['mean_test_score'][candidate],
                results['std_test_score'][candidate]))
            print("Parameters: {0}".format(results['params'][candidate]))
            print("")
```

0.1 SVM Regressor

0.2 Medida de erro

In [12]:

```
list_exp_C = range(-5, 15, 5)
list_exp_gamma = range(-15, 3, 3)
list_epsilon = np.arange(0.05, 1, 0.5)

#Lista de parametros para utilizar nos demais
C_list = []
gamma_list = []

best_params = []
best_score = 0
mse = 0
for exp_gamma in list_exp_gamma:
    gamma = 2**exp_gamma
    gamma_list.append(gamma)
    for exp_C in list_exp_C:
        C = 2**exp_C
        C_list.append(C)
        for epsilon in list_epsilon:
            clf = SVR(gamma=gamma, C=C, epsilon=epsilon)
            clf.fit(x_train, y_train)
            score = clf.score(x_test, y_test)
            if score > best_score:
                best_score = score
                best_params = [gamma, C, epsilon]
                mse = mean_squared_error(clf.predict(x_test), y_test)
print("Best Score: ", best_score, " with params: ", best_params, " and MSE: ", mse)

C_list = np.unique(C_list)
gamma_list = np.unique(gamma_list)
list_epsilon = np.unique(list_epsilon)
```

Best Score: 0.7101346849881802 with params: [0.000244140625, 1024, 0.05] and MSE: 19.45487953760004

1 Random search

In [64]:

```
param_dist = {"C": [2**-5, 2**0, 2**5, 2**10, 2**15],
              "gamma": [2**-15, 2**-10, 2**-5, 2**0, 2**5],
              "epsilon": [0.05, 0.1, 0.5, 0.8, 1.0]}

clf = SVR()

random_search = RandomizedSearchCV(clf, param_distributions=param_dist, cv=4)

random_search.fit(x_train, y_train)

report(random_search.cv_results_)
```

Model with rank: 1
Mean validation score: 0.780 (std: 0.045)
Parameters: {'gamma': 3.0517578125e-05, 'epsilon': 0.5, 'C': 32768}

Model with rank: 2
Mean validation score: 0.725 (std: 0.042)
Parameters: {'gamma': 3.0517578125e-05, 'epsilon': 0.5, 'C': 1024}

Model with rank: 3
Mean validation score: 0.723 (std: 0.040)
Parameters: {'gamma': 3.0517578125e-05, 'epsilon': 1.0, 'C': 1024}

2 Grid search

In [63]:

```
grid_search = GridSearchCV(clf, param_grid=param_dist, cv=5)

grid_search.fit(x_train, y_train)

report(grid_search.cv_results_)
```

/home/pazeto/anaconda3/envs/M0431A/lib/python3.7/site-packages/sklearn/model_selection/_search.py:841: DeprecationWarning: The default of the `iid` parameter will change from True to False in version 0.22 and will be removed in 0.24. This will change numeric results when test-set sizes are unequal.
DeprecationWarning)

Model with rank: 1
Mean validation score: 0.746 (std: 0.040)
Parameters: {'C': 32768, 'epsilon': 1.0, 'gamma': 3.0517578125e-05}

Model with rank: 2
Mean validation score: 0.744 (std: 0.039)
Parameters: {'C': 32768, 'epsilon': 0.5, 'gamma': 3.0517578125e-05}

Model with rank: 3
Mean validation score: 0.744 (std: 0.034)
Parameters: {'C': 32768, 'epsilon': 0.1, 'gamma': 3.0517578125e-05}

3 Otimização bayesiana

In [4]:

```
# !pip install hyperopt
from hyperopt import hp, tpe, fmin, space_eval

def my_svr_fun(params):
    C, gamma, epsilon = params['C'], params['gamma'], params['epsilon']
    clf = SVR(gamma=gamma, C=C, epsilon=epsilon)
    clf.fit(x_train, y_train)
    score = clf.score(x_test, y_test)
    return -score

param_hyperopt = {
    'C': 2 ** hp.uniform('C', -5, 15),
    'gamma': 2 ** hp.uniform('gamma', -15, 3),
    'epsilon': hp.uniform('epsilon', 0.05, 1),
}

best_params = fmin(fn = my_svr_fun,
                   space = param_hyperopt, algo=tpe.suggest,
                   max_evals = 125)

print("Melhores paramtros: ", space_eval(param_hyperopt, best_params))
print("Score: ", my_svr_fun(space_eval(param_hyperopt, best_params))*-1)
```

```
100%|██████████| 125/125 [00:50<00:00, 2.48it/s, best loss: -0.825205
8541104953]
Melhores paramtros: {'C': 18254.13344755772, 'epsilon': 0.15646708299
651826, 'gamma': 3.194344320388373e-05}
Score: 0.8252058541104953
```

4 PSO

In [160]:

```
from pyswarm import pso

#primeiro elemento é o mais baixo e mais alte de C, segundo de gamma e terceiro de
lb = [2** -5, 2** -15, 0.05]
ub = [2**15, 2**3, 1.0]

#dado que o pso vai esperar o menor valor e
# queremos o melhor score devemos negativar o score, o mais baixo terá melhores par
def my_svr_fun(params):
    # print(params)
    C, gamma, epilson = params
    clf = SVR(gamma=gamma, C=C, epsilon=epilson)
    clf.fit(x_train, y_train)
    score = clf.score(x_test, y_test)
    return -score

best_params, best_score = pso(my_svr_fun, lb, ub, swarmsize=11, maxiter=11)
print("Parametros: ", best_params)
print("Score:", best_score*-1)
```

```
Stopping search: maximum iterations reached --> 11
Parametros: [1.86712162e+04 3.05175781e-05 5.29600172e-02]
Score: 0.8296045658080052
```

5 Simulated annealing

In []:

```
from simanneal import Annealer
import random

class SearchBestParamsSVR(Annealer):

    def move(self):

        """permuta a lista de parametros possíveis"""

        random.shuffle(self.state['C'])
        random.shuffle(self.state['gamma'])
        random.shuffle(self.state['epsilon'])

    def energy(self):
        """Calcula o melhor score"""

        C, gamma, epsilon = self.state['C'][0], self.state['gamma'][0], self.state['epsilon'][0]
        clf = SVR(gamma=gamma, C=C, epsilon=epsilon)
        clf.fit(x_train, y_train)
        score = clf.score(x_test, y_test)

        return -score

param_dist = {"C": [2**-5, 2**0, 2**5, 2**10, 2**15],
              "gamma": [2**-15, 2**-10, 2**-5, 2**0, 2**5],
              "epsilon": [0.05, 0.1, 0.5, 0.8, 1.0]}

tsp = SearchBestParamsSVR(param_dist)

params, score = tsp.anneal()
```

In [119]:

```
print("Parametros C gamma e epsilon:", params['C'][0], params['gamma'][0], params['epsilon'][0])
print("Score:", score*-1)
```

Parametros C gamma e epsilon: 32768 3.0517578125e-05 0.05
Score: 0.8220660478366459

6 CMA-ES

In [73]:

```
# !pip install cma
import cma
# help(cma.fmin)
def my_svr_fun(params):
    C, gamma, epilson = params
    clf = SVR(gamma=gamma, C=C, epsilon=epilson)
    clf.fit(x_train, y_train)
    score = clf.score(x_test, y_test)
    return -score

lb = [2**-5, 2**-15, 0.05]
ub = [2**15, 2**3, 1]

opt = { 'bounds': [lb, ub], 'verb_disp':1000}
opt['scaling_of_variables'] = [2**10, 2**1, 0.01]

es = cma.fmin(my_svr_fun, [1, 1, 1], 2.5, opt)

print("Parametros C gamma e epilson: ", es[0])
print("Score: ", es[1]*-1)
```

(3_w,7)-aCMA-ES (mu_w=2.3,w_l=58%) in dimension 3 (seed=769965, Wed May 1 11:49:26 2019)

	Iterat	#Fevals	function value	axis ratio	sigma	min&max	std	t[m:s]
	1	7	-1.838206063591308e-04	1.0e+00	2.08e+00	2e+00	2e+00	0:0
0.1	2	14	1.480056132296959e-04	1.4e+00	1.93e+00	1e+00	2e+00	0:0
0.2	3	21	-9.133457200733330e-03	1.8e+00	1.83e+00	1e+00	2e+00	0:0
0.3	22	154	-6.326321821002134e-01	1.2e+01	1.54e-01	1e-02	1e-01	0:0
3.5	34	238	-6.475068819370783e-01	4.6e+01	7.24e-02	1e-03	6e-02	0:0
7.6	46	322	-7.111023370763685e-01	3.3e+02	1.39e-01	1e-03	2e-01	0:1
3.1	53	371	-7.681652760797661e-01	9.7e+02	7.36e-01	4e-03	2e+00	0:1
9.7	56	392	-7.774315533913859e-01	1.1e+03	9.40e-01	3e-03	2e+00	0:2
7.1	60	420	-7.824298371973549e-01	1.0e+03	8.63e-01	2e-03	1e+00	0:3
6.7	64	448	-7.975142264359517e-01	1.2e+03	6.91e-01	1e-03	1e+00	0:4
8.6	67	469	-7.968387811299247e-01	1.3e+03	8.05e-01	9e-04	2e+00	1:0
1.2	70	490	-8.010360811367336e-01	1.8e+03	7.51e-01	7e-04	1e+00	1:1
3.6	74	518	-7.956126969487143e-01	1.6e+03	9.39e-01	1e-03	1e+00	1:2
9.4	78	546	-8.042409170371525e-01	1.6e+03	1.31e+00	2e-03	2e+00	1:4
6.8	81	567	-7.969227474040904e-01	1.7e+03	1.28e+00	1e-03	1e+00	2:0
4.8	84	588	-8.013909088537767e-01	2.0e+03	1.39e+00	1e-03	2e+00	2:2
5.1	88	616	-8.018597706470010e-01	1.5e+03	1.52e+00	1e-03	1e+00	2:4

466	3262	-8.256966351184095e-01	2.6e+02	6.48e-02	3e-11	7e-05	53:
07.0							
473	3311	-8.256993397236930e-01	2.3e+02	7.05e-02	3e-11	6e-05	54:
25.5							
482	3374	-8.256917416376404e-01	1.1e+02	1.13e-01	7e-11	5e-05	55:
50.6							
490	3430	-8.256894822402588e-01	9.9e+01	9.33e-02	4e-11	4e-05	57:
10.6							
498	3486	-8.257054469894023e-01	7.9e+01	1.11e-01	5e-11	3e-05	58:
38.8							
508	3556	-8.257005453302735e-01	9.8e+01	1.53e-01	5e-11	4e-05	60:
00.4							
519	3633	-8.256941680231893e-01	2.3e+02	2.29e-01	6e-11	7e-05	61:
22.9							
530	3710	-8.257020333945037e-01	2.1e+02	1.43e-01	3e-11	3e-05	62:
47.5							
541	3787	-8.256941719645042e-01	3.0e+02	2.00e-01	3e-11	7e-05	64:
13.0							
553	3871	-8.257005250102156e-01	3.4e+02	2.49e-01	4e-11	5e-05	65:
44.5							
565	3955	-8.257050357771927e-01	4.4e+02	1.10e-01	1e-11	2e-05	67:
17.8							
577	4039	-8.257007381794392e-01	1.4e+03	4.05e-02	4e-12	8e-06	68:
49.3							
589	4123	-8.256982696352093e-01	1.1e+03	4.68e-02	4e-12	7e-06	70:
18.8							
601	4207	-8.256983777314512e-01	2.1e+03	6.43e-02	7e-12	1e-05	71:
51.3							
613	4291	-8.256978744085418e-01	3.4e+03	7.51e-02	8e-12	1e-05	73:
23.3							
625	4375	-8.256909369531588e-01	4.4e+03	6.49e-02	5e-12	8e-06	74:
55.1							
638	4466	-8.257026310997099e-01	2.9e+03	1.03e-01	4e-12	7e-06	76:
32.5							
650	4550	-8.256973709038034e-01	4.3e+03	6.20e-02	2e-12	4e-06	78:
01.5							

termination on tolstagnation=174 (Wed May 1 13:07:30 2019)

final/bestever f-value = -8.257023e-01 -8.257230e-01

incumbent solution: [19397.06710389928, 3.051757888265911e-05, 0.3603345945951138]

std deviation: [0.0005189705279846922, 4.266018454044987e-12, 3.936943727446785e-08]

Parametros C gamma e epilson: [1.94095292e+04 3.05266127e-05 3.60585058e-01]

Score: 0.8257229759217513