

プログラム

実行環境と用いた言語・ライブラリを以下の表 1 に示す。

表 1: プログラムの実行環境

OS	: Microsoft Windows 10 Pro (64bit)
CPU	: Intel(R) Core(TM) i5-4300U
RAM	: 4.00 GB
使用言語	: Python3.6
可視化	: matplotlib ライブラリ

Listings 1: assignment2.py

```
1  # -*- coding: utf-8 -*-
2
3
4  import numpy as np
5  import matplotlib.pyplot as plt
6
7
8  def true_model(x):
9      pix = np.pi * x
10     target = np.sin(pix) / pix + 0.1 * x
11     return target
12
13
14  def gauss_kernel(x, c, h):
15     return np.exp(-(x - c)**2 / (2*h**2))
16
17
18  def generate_sample(xmin, xmax, sample_size):
19     x = np.linspace(start=xmin, stop=xmax, num=sample_size)
20     target = true_model(x)
21     noise = 0.05 * np.random.normal(loc=0., scale=1., size=sample_size)
22     return x, target + noise
23
24
25  def split(x, y, n_split=5):
26     n_data = len(y)
27     n_data_in_one_split = int(n_data / n_split)
28     idx = np.arange(n_data)
29     np.random.shuffle(idx)
30
31     x_split = []
32     y_split = []
33     for i in range(n_split):
```

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34         idx_start = i * n_data_in_one_split
35         idx_end = (i+1) * n_data_in_one_split
36         if idx_end == n_data:
37             idx_end = None
38         x_split.append(x[idx_start:idx_end])
39         y_split.append(y[idx_start:idx_end])
40     return x_split, y_split
41
42
43 def split_train_test(x_split, y_split, k):
44     n_split = len(y_split)
45     x_test, y_test = x_split[k], y_split[k]
46     x_train, y_train = [], []
47     for _k in range(n_split):
48         if _k != k:
49             x_train.extend(x_split[_k])
50             y_train.extend(y_split[_k])
51     x_train = np.array(x_train)
52     y_train = np.array(y_train)
53     return x_train, y_train, x_test, y_test
54
55
56 def calc_design_matrix(x, c, h, kernel):
57     return kernel(x[None], c[:, None], h)
58
59
60 def solve_gauss_kernel_model(x, y, h, lamb):
61     k = calc_design_matrix(x, x, h, gauss_kernel)
62     theta = np.linalg.solve(
63         k.T.dot(k) + lamb*np.identity(len(k)),
64         k.T.dot(y[:, None]),
65     )
66     return theta
67
68
69 def solve_gauss_kernel_sparse_model(x, y, h, lamb, eps=1e-2, iter_max=100,
70     n_look=5):
71     def update(theta, z, u, lamb, K, y):
72         theta = np.linalg.inv(K.T.dot(K) + np.eye(len(K))).dot(K.dot(y[:,
73             None]) + z - u)
74
75         z_plus = theta + u - lamb
76         z_plus[z_plus < 0] = 0
77         z_minus = theta + u + lamb
78         z_minus[z_minus > 0] = 0
79         z = z_plus + z_minus

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79         u = u + theta - z
80         return theta, z, u
81
82     def compute_lagrange_func(theta, z, u, lamb, K, y):
83         loss = (
84             (1/2) * np.linalg.norm(K.dot(theta) - y)
85             + lamb * np.linalg.norm(z, ord=1)
86             + u.T.dot(theta - z)
87             + (1/2) * np.linalg.norm(theta - z)
88             )
89         return loss
90
91     K = calc_design_matrix(x, x, h, gauss_kernel)
92     theta = np.linalg.solve(K.T.dot(K), K.T.dot(y[:, None]))
93     z = np.random.rand(*theta.shape) * 0.1
94     u = np.random.rand(*theta.shape) * 0.1
95     lagrange_list = []
96     for i in range(iter_max):
97         theta, z, u = update(theta, z, u, lamb, K, y)
98         lagrange = compute_lagrange_func(theta, z, u, lamb, K, y)
99         if i < n_look:
100             lagrange_list.append(lagrange)
101         else:
102             lagrange_list = lagrange_list[1:] + [lagrange]
103             if max(lagrange_list) - min(lagrange_list) < eps:
104                 break
105     return theta, z, u, i+1
106
107
108 def compute_loss(x_train, x_test, y, h, theta, lamb):
109     k = calc_design_matrix(x_train, x_test, h, gauss_kernel)
110     loss = (1/2)*np.linalg.norm(k.dot(theta) - y)
111     # loss += (lamb/2)*np.linalg.norm(theta)
112     return loss
113
114
115 def main():
116     np.random.seed(0) # set the random seed for reproducibility
117
118     # create sample
119     xmin, xmax = -3, 3
120     sample_size = 50
121     n_split = 5
122     x, y = generate_sample(xmin=xmin, xmax=xmax, sample_size=sample_size)
123     # print(x.shape, y.shape)
124
125     x_split, y_split = split(x, y, n_split=n_split)

```

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126     # print(x_split[0].shape, y_split[0].shape)
127
128     # global search
129     h_cands = [1e-2, 1e-1, 1, 1e1,]
130     lamb_cands = [1e-6, 1e-5, 1e-4, 1e-3, 1e-2, 1e-1,]
131
132     # local search
133     #searched_range_base = np.arange(0.5, 1.5, 0.1)
134     #h_cands = 1.0 * searched_range_base
135     #lamb_cands = 1e-6 * searched_range_base
136
137     loss_min = 1e8
138     h_best = None
139     lamb_best = None
140     theta_best = None
141     n_iter_best = None
142
143     n_row = len(lamb_cands)
144     n_col = len(h_cands)
145     fig = plt.figure(figsize=(n_col*4, n_row*4))
146     fig_idx = 0
147
148     for lamb in lamb_cands:
149         for h in h_cands:
150             losses = []
151             for k in range(n_split):
152                 x_train, y_train, x_test, y_test = split_train_test(x_split,
153 y_split, k)
154
155                 # print(x_train.shape, y_train.shape)
156
157                 theta, z, u, n_iter = solve_gauss_kernel_sparse_model(
158                     x_train, y_train, h, lamb,
159                     eps=1e-3, iter_max=200, n_look=10,
160                 )
161                 loss_k = compute_loss(x_train, x_test, y_test, h, theta,
162 lamb)
163
164                 losses.append(loss_k)
165             loss = np.mean(losses)
166
167             if loss < loss_min:
168                 loss_min = loss
169                 h_best = h
170                 lamb_best = lamb
171                 theta_best = theta
172                 n_iter_best = n_iter
173
174         # for visualization

```

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171         X = np.linspace(start=xmin, stop=xmax, num=5000)
172         true = true_model(X)
173         K = calc_design_matrix(x_train, X, h, gauss_kernel)
174         prediction = K.dot(theta)
175
176         # visualization
177         fig_idx += 1
178         ax = fig.add_subplot(n_row, n_col, fig_idx)
179         ax.set_title('$h = {}, \backslash \lambda = {}$, L = {:.2f}'.format(h,
180 lamb, loss))
181         ax.scatter(x, y, c='green', marker='o', label='data')
182         ax.plot(X, true, linestyle='dashed', label='true')
183         ax.plot(X, prediction, linestyle='solid', label='predicted')
184         ax.legend()
185
186     print('Best Model')
187     print('\th = {}'.format(h_best))
188     print('\tlambda = {}'.format(lamb_best))
189     print('\tloss = {}'.format(loss_min))
190     print('\tn_iter: {}'.format(n_iter_best))
191     print('\ttheta: \n', theta_best)
192
193     plt.savefig('../figures/assignment2_result.png')
194     plt.show()
195
196 if __name__ == '__main__':
197     main()

```