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1 import argparse
2 from pathlib import Path
3 from collections import defaultdict
4 from datetime import datetime
5
6 import numpy as np
7
8 from my_logger import get_my_logger
9 from utils import load_data, Trainer, plot_accs, tr_va_split
10 from preprocess import ImagePreprocessor, TextPreprocessor
11 from save_csv import results_to_csv
12
13 if __name__ == "__main__":
14     # logger
15     logger = get_my_logger(Path(__file__).name)
16     logger.info('--- [START {0:%Y-%m-%d_%H-%M-%S}]
17 {1}'.format(datetime.now(), '-' * 100))
18
19     parser = argparse.ArgumentParser(description='cs289 hw1')
20     parser.add_argument('problem', type=int, help='problem number (e.g. 2, 3,
21 4)')
22     parser.add_argument('data_name', type=str, help='dataset name (e.g.
23 mnist, spam, cifar10)')
24     args = parser.parse_args()
25
26     # -- configuration -----
27     data_name = args.data_name # cifar10 mnist spam
28
29     split_and_save = 0
30     preprocess = 0
31     tune_n_tr = 0
32     tune_hparameter = 0
33     train_and_predict = 0
34     split = 'holdout' # kfold holdout
35     n_fold = 1
36     kernel = 'linear'
37
38     n_trs = {
39         'mnist': None, # 5000
40         'spam': None,
41         'cifar10': None # 5000
42     }
43
44     if args.problem == 2:
45         split_and_save = 1
46     elif args.problem == 3:
47         tune_n_tr = 1
48     elif args.problem == 4:
49         tune_hparameter = 1
50         n_trs['mnist'] = 10000
51     elif args.problem == 5:
52         tune_hparameter = 1
53         split = 'kfold'
54         n_fold = 5
55     elif args.problem == 6:
56         preprocess = 1
57         train_and_predict = 1
58         split = 'kfold'
59         n_fold = 5
60         kernel = 'rbf'

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58
59 logger.info(f'data:\t{data_name}')
60 logger.info(f'prepro:\t{preprocess}')
61 logger.info(f'split:\t{split}')
62 logger.info(f'n_fold:\t{n_fold}')
63
64 # seed
65 seed = 189
66 np.random.seed(seed)
67 logger.info(f'seed:\t{seed}')
68
69 n_vas = {
70     'mnist': 10000,
71     'spam': 0.2,
72     'cifar10': 5000
73 }
74 n_va = n_vas[data_name]
75
76 preprocessors = {
77     'mnist': ImagePreprocessor(
78         normalize=False, hog_tf=True, pca_tf=False, lbp_tf=False,
79         win_size=(28, 28), block_size=(8, 8),
80         block_stride=(4, 4), cell_size=(8, 8), nbins=9
81     ),
82     'spam': TextPreprocessor(),
83     'cifar10': ImagePreprocessor(
84         normalize=False, hog_tf=False, pca_tf=True, lbp_tf=False
85     )
86 }
87 preprocessor = preprocessors[data_name]
88
89 n_tr_spaces = {
90     'mnist': [100, 200, 500, 1000, 2000, 5000, 10000],
91     'spam': [100, 200, 500, 1000, 2000, 4138],
92     'cifar10': [100, 200, 500, 1000, 2000, 5000]
93 }
94 n_tr_space = n_tr_spaces[data_name]
95
96 all_params = {
97     'mnist': {
98         'C': 2e1,
99         'kernel': kernel,
100         'gamma': 'scale',
101         'random_state': seed
102     },
103     'spam': {
104         'C': 1e1,
105         'kernel': kernel,
106         'gamma': 'scale',
107         'random_state': seed
108     },
109     'cifar10': {
110         'C': 1e1,
111         'kernel': kernel,
112         'gamma': 'scale',
113         'random_state': seed
114     }
115 }
116 params = all_params[data_name]

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117 all_parameter_spaces = {
118     'mnist': {
119         'C': np.logspace(-10, 10, 11, base=10)
120     },
121     'spam': {
122         'C': np.logspace(-10, 4, 8, base=10)
123     },
124     'cifar10': {
125         'C': np.logspace(-10, 10, 11, base=10)
126     }
127 }
128 parameter_spaces = all_parameter_spaces[data_name]
129 # -----
130
131 # load data
132 x_train, x_test, y_train = load_data(data_name)
133 logger.info(f'x_train:\t{x_train.shape}')
134 logger.info(f'y_train:\t{y_train.shape}')
135 logger.info(f'x_test:\t{x_test.shape}')
136
137 # preprocessing
138 if preprocess:
139     logger.info('** preprocess **')
140     x_train, x_test = preprocessor.scale(x_train, x_test)
141     logger.info(f'x_train:\t{x_train.shape}')
142     logger.info(f'x_test:\t{x_test.shape}')
143
144 # tune training data size
145 if tune_n_tr:
146     logger.info('** tune training data size**')
147     accs = defaultdict(dict)
148     for n_tr in n_tr_space:
149         trainer = Trainer(params, x_train, x_test, y_train,
verbose=False)
150         trainer.validate(split, n_fold=n_fold, n_va=n_va, n_tr=n_tr,
shuffle=True)
151
152         tr_acc = trainer.scores_mean['tr_acc']
153         va_acc = trainer.scores_mean['va_acc']
154         accs['train'][n_tr] = tr_acc
155         accs['valid'][n_tr] = va_acc
156         logger.info(f'[{n_tr}]\tttr:{tr_acc:.5f}\tva:{va_acc:.5f}')
157
158         path = Path(f'figures/{data_name}_acc_vs_n_tr.png')
159         xlabel = 'Training data size'
160         xlim = (0, max(n_tr_space) * 1.1)
161         plot_accs(accs, data_name, path, xlabel, xlim, False)
162
163 # tune C of SVM with holdout
164 if tune_hparameter:
165     logger.info('** tune hyper parameter **')
166     n_tr = n_trs[data_name]
167     best_acc = 0
168     best_params = {}
169
170     for parameter, space in parameter_spaces.items():
171         accs = defaultdict(dict)
172         for value in space:
173             params[parameter] = value

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174         trainer = Trainer(params, x_train, x_test, y_train,
verbose=False)
175         trainer.validate(split, n_fold=n_fold, n_va=n_va, n_tr=n_tr,
shuffle=True)
176
177         tr_acc = trainer.scores_mean['tr_acc']
178         va_acc = trainer.scores_mean['va_acc']
179         accs['train'][value] = tr_acc
180         accs['valid'][value] = va_acc
181         logger.info(f'[{parameter}]:{value:.0e}]\ttr:{tr_acc:.5f}\tva:
{va_acc:.5f}')
182
183         if va_acc > best_acc:
184             best_params = params.copy()
185             best_acc = va_acc
186
187         path = Path(f'figures/{data_name}_acc_vs_{parameter}.png')
188         xlabel = parameter
189         xlim = (min(space) * 0.9, max(space) * 1.1)
190         plot_accs(accs, data_name, path, xlabel, xlim, log_scale=True)
191
192         params = best_params.copy()
193         logger.info(f'[best_acc]\t{best_acc:.5f}')
194         logger.info(f'[best_params]\t{best_params}')
195
196     # split and save
197     if split_and_save:
198         logger.info('** split and save **')
199         n_tr = n_trs[data_name]
200         cv = tr_va_split(x_train, x_test, n_va, shuffle=True)
201         ind_tr, ind_va = next(cv)
202         x_tr, x_va = x_train[ind_tr:n_tr], x_train[ind_va]
203         y_tr, y_va = y_train[ind_tr:n_tr], y_train[ind_va]
204
205         logger.info(f'x_tr: {x_tr.shape}')
206         logger.info(f'x_va: {x_va.shape}')
207         logger.info(f'y_tr: {y_tr.shape}')
208         logger.info(f'y_va: {y_va.shape}')
209
210         save_dir = Path('data', 'split')
211         np.save(Path(save_dir, f'{data_name}_x_tr.npy'), x_tr)
212         np.save(Path(save_dir, f'{data_name}_x_va.npy'), x_va)
213         np.save(Path(save_dir, f'{data_name}_y_tr.npy'), y_tr)
214         np.save(Path(save_dir, f'{data_name}_y_va.npy'), y_va)
215
216     # train and predict
217     if train_and_predict:
218         logger.info('** train and predict **')
219         n_tr = n_trs[data_name]
220
221         trainer = Trainer(params, x_train, x_test, y_train, verbose=True)
222         trainer.validate(split, n_fold=n_fold, n_va=n_va, n_tr=n_tr,
shuffle=True)
223
224         tr_acc = trainer.scores_mean['tr_acc']
225         va_acc = trainer.scores_mean['va_acc']
226         logger.info(f'[all]\ttr:{tr_acc:.5f}\tva:{va_acc:.5f}')
227
228         submit_path =
Path(f'data/output/submit_{data_name}_{va_acc:.5f}.csv', index=False)

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results_to_csv(trainer.y_test_pred, submit_path)
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