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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.
mnist, spam, cifar10)')
23     args = parser.parse_args()
24
25     # -- configuration -----
26     data_name = args.data_name # cifar10 mnist spam
27
28     split_and_save = 0
29     preprocess = 0
30     tune_n_tr = 0
31     tune_hparameter = 0
32     train_and_predict = 0
33     split = 'holdout' # kfold holdout
34     n_fold = 1
35     kernel = 'linear'
36
37     n_trs = {
38         'mnist': None, # 5000
39         'spam': None,
40         'cifar10': None # 5000
41     }
42
43     if args.problem == 2:
44         split_and_save = 1
45     elif args.problem == 3:
46         tune_n_tr = 1
47     elif args.problem == 4:
48         tune_hparameter = 1
49         n_trs['mnist'] = 10000
50     elif args.problem == 5:
51         tune_hparameter = 1
52         split = 'kfold'
53         n_fold = 5
54     elif args.problem == 6:
55         preprocess = 1
56         train_and_predict = 1
57         split = 'kfold'
58         n_fold = 5
59         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}]\ttr:{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     verbose=False)
175     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,
223                      shuffle=True)
224
225     tr_acc = trainer.scores_mean['tr_acc']
226     va_acc = trainer.scores_mean['va_acc']
227     logger.info(f'all\ttr:{tr_acc:.5f}\tva:{va_acc:.5f}')
228
229     submit_path =
Path(f'data/output/submit_{data_name}_{va_acc:.5f}.csv', index=False)

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```
229 |     results_to_csv(trainer.y_test_pred, submit_path)
230 | 
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