

Implement a new Neural Network Cross Validation

The BRAPH 2 Developers

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This is the developer tutorial for implementing a new neural network cross validation. In this Tutorial, we will explain how to create the generator file `*.gen.m` for a new neural network cross validation, which can then be compiled by `braph2genesis`. All kinds of neural network cross validation are (direct or indirect) extensions of the base element `NNCrossValidation`. Here, we will use as examples the cross validation for multi-layer perceptron regressors `NNRegressorMLP_CrossValidation` and the cross validation for multi-layer perceptron classifiers `NNClassifierMLP_CrossValidation`.

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Implementation of the Cross Validation for Regressors

We will start by implementing in detail `NNRegressorMLP_CrossValidation`, which is a direct extension of `NNCrossValidation`. A cross validation for multi-layer perceptron regressors (`NNRegressorMLP_CrossValidation`) is a procedure designed for evaluating multi-layer perceptron regressors using cross-validation.

Code 1: `NNRegressorMLP_CrossValidation` element

header. The header section of the generator code for `_NNRegressorMLP_CrossValidation.gen.m` provides the general information about the `NNRegressorMLP_CrossValidation` element.

```

1 %% iheader!
2 NNRegressorMLP_CrossValidation < NNCrossValidation (nncv, neural network
   cross-validation) is a process for evaluating multi-layer perceptron
   regressors using cross-validation. ①
3
4 %%% idescription!
5 A cross validation for multi-layer perceptron regressors (
   NNRegressorMLP_CrossValidation) is a process that facilitates the
   evaluation of multi-layer perceptron regressors using cross-validation.
6 It involves splitting a dataset into multiple subsets (folds), training the
   model on some folds while validating on others, and then repeating the
   process for all combinations of folds.
7 This helps in assessing the generalization performance of the model and
   detecting overfitting.
8
9 To train all the neural networks for all folds, use: nncv.get('TRAIN')
```

① defines `NNRegressorMLP_CrossValidation.gen` as a subclass of `NNCrossValidation`. The moniker will be `nncv`.

Code 2: `NNRegressorMLP_CrossValidation` element prop

update. The `props_update` section of the generator code for `_NNRegressorMLP_CrossValidation.gen.m` updates the properties of the `NNRegressorMLP_CrossValidation` element. This defines the core properties of the data point.

```

1 %% iprops_update!
2
3 %%% iprop!
4 NAME (constant, string) is the name of the cross-validation.
5 %%% idefault!
6 'NNRegressorMLP_CrossValidation'
7
8 %%% iprop!
9 DESCRIPTION (constant, string) is the description of the cross-validation.
10 %%% idefault!
11 'A cross validation for multi-layer perceptron regressors (
   NNRegressorMLP_CrossValidation) is a process that facilitates the
   evaluation of multi-layer perceptron regressors using cross-validation.
   It involves splitting a dataset into multiple subsets (folds),
   training the model on some folds while validating on others, and then
   repeating the process for all combinations of folds. This helps in
   assessing the generalization performance of the model and detecting
   overfitting.'
```

```

12
13 %%% iprop!
```

```

14 TEMPLATE (parameter, item) is the template of the neural cross-validation.
15 %%% isettings!
16 'NNRegressorMLP_CrossValidation'
17
18 %%% iprop!
19 ID (data, string) is a few-letter code for the cross-validation.
20 %%% idefault!
21 'NNRegressorMLP_CrossValidation ID'
22
23 %%% iprop!
24 LABEL (metadata, string) is an extended label of the cross-validation.
25 %%% idefault!
26 'NNRegressorMLP_CrossValidation label'
27
28 %%% iprop!
29 NOTES (metadata, string) are some specific notes about the cross-validation.
30 %%% idefault!
31 'NNRegressorMLP_CrossValidation notes'
32
33 %%% iprop!
34 NN_TEMPLATE (parameter, item) is the neural network template to set all
    neural network parameters.
35 %%% isettings!
36 'NNRegressorMLP' ①
37
38 %%% iprop!
39 NNEVALUATOR_TEMPLATE (parameter, item) is the neural network evaluator
    template to set all evaluator parameters.
40 %%% isettings!
41 'NNRegressorMLP_Evaluator' ②
42
43 %%% iprop!
44 NN_LIST (result, itemlist) contains the neural network models corresponding
    to k folds.
45 %%% icalculate!
46 d_list = nncv.get('D_LIST');
47
48 if isempty(d_list)
49     value = {};
50 else
51     for i = 1:length(d_list) ③
52         d_training_set{i} = d_list;
53         d_training_set{i}(i) = []; % Exclude the i-th element
54         d_training_set{i} = NNDatasetCombine('D_LIST', d_training_set{i}).
            get('D');
55     end
56
57     d_training_set = d_training_set';
58
59     if ~isa(nncv.get('NN_TEMPLATE'), 'NoValue')
60         nn_template = nncv.get('NN_TEMPLATE'); ④
61     else
62         nn_template = NNRegressorMLP( ... ⑤
63             'EPOCHS', nncv.get('EPOCHS'), ...
64             'BATCH', nncv.get('BATCH'), ...
65             'SHUFFLE', nncv.get('SHUFFLE'), ...
66             'SOLVER', nncv.get('SOLVER'), ...
67             'VERBOSE', nncv.get('VERBOSE'), ...
68             'PLOT_TRAINING', nncv.get('PLOT_TRAINING'));
69     end

```

① defines the neural network template to be NNRegressorMLP, which will be used to set the parameters for all the neural network regressors for the cross validation.

② defines the neural network evaluator template to be NNRegressorMLP_Evaluator, which will be used to set the parameters for all the neural network evaluator for the cross validation.

③ constructs the training sets iteratively by concatenating all of the remaining folds after excluding the one as the validation set.

④ and ⑤ set the parameters to all the NNRegressorMLP based on either the NN_TEMPLATE or the parameters from this cross validation.

```

70
71     value = cellfun(@(d) NNRegressorMLP( ... ⑥
72         'TEMPLATE', nn_template, 'D', d), ...
73         d_training_set, 'UniformOutput', false);
74 end
75
76 %%% iprop!
77 EVALUATOR_LIST (result, itemlist) contains the evaluators corresponding to k
    folds.
78 %%% icalculate!
79 %%% iprop!
80 EVALUATOR_LIST (result, itemlist) contains the evaluators corresponding to k
    folds.
81 %%% icalculate!
82 d_list = nncv.get('D_LIST');
83 nn_list = nncv.get('NN_LIST');
84
85 if ~isa(nncv.get('NNEVALUATOR_TEMPLATE'), 'NoValue')
86     nne_template = nncv.get('NNEVALUATOR_TEMPLATE'); ⑦
87 else
88     nne_template = NNRegressorMLP_Evaluator( ... ⑧
89         'P', nncv.get('P'));
90 end
91
92 value = cellfun(@(d, nn) NNRegressorMLP_Evaluator('TEMPLATE', nne_template,
93     'D', d, 'NN', nn), ... ⑨
    d_list, nn_list, 'UniformOutput', false);

```

⑥ initializes all of the NNRegressorMLPs.

⑦ and ⑧ set the parameters to all the NNRegressorMLP_Evaluator based on either the NNEVALUATOR_TEMPLATE or the parameters from this cross validation.

⑨ initializes all of the NNRegressorMLP_Evaluators.

Code 3: NNRegressorMLP_CrossValidation element props. The props section of generator code for `_NNRegressorMLP_CrossValidation.gen.m` defines the properties to be used in `NNRegressorMLP_CrossValidation`.

```

1 %%% iprops!
2
3 %%% iprop!
4 P (parameter, scalar) is the permutation number. ①
5 %%% idefault!
6 1e+2
7 %%% icalculate!
8 check = value > 0 && value == round(value);
9
10 %%% iprop!
11 AV_CORR (result, rvector) provides the metric of the correlation of
    coefficients. ②
12 %%% icalculate!
13 e_list = nncv.get('EVALUATOR_LIST');
14
15 value = cellfun(@(e) e.get('CORR'), ...
16     e_list, 'UniformOutput', false);
17
18 if isempty(value)
19     value = [];
20 else
21     value = mean(cell2mat(value), 1);
22 end
23
24 %%% iprop!

```

① defines the number for permutation feature importance.

②, ③, ④, and ⑤ calculates the average metrics, including the correlation of coefficients, the coefficient of determination, the mean absolute error, the mean squared error, and the root mean squared error.

```

25 AV_DET (result, rvector) provides the coefficient of determination, a
    measure showing how well the predictions are replicated by the model.
    ③
26 %%% icalculate!
27 e_list = nncv.get('EVALUATOR_LIST');
28
29 value = cellfun(@(e) e.get('DET'), ...
30     e_list, 'UniformOutput', false);
31
32 if isempty(value)
33     value = [];
34 else
35     value = mean(cell2mat(value), 1);
36 end
37
38 %%% iprop!
39 AV_MAE (result, rvector) provides the metric of the mean absolute error. ④
40 %%% icalculate!
41 e_list = nncv.get('EVALUATOR_LIST');
42
43 value = cellfun(@(e) e.get('MAE'), ...
44     e_list, 'UniformOutput', false);
45
46 if isempty(value)
47     value = [];
48 else
49     value = mean(cell2mat(value), 1);
50 end
51
52 %%% iprop!
53 AV_MSE (result, rvector) provides the metric of the mean squared error. ⑤
54 %%% icalculate!
55 e_list = nncv.get('EVALUATOR_LIST');
56
57 value = cellfun(@(e) e.get('MSE'), ...
58     e_list, 'UniformOutput', false);
59
60 if isempty(value)
61     value = [];
62 else
63     value = mean(cell2mat(value), 1);
64 end
65
66 %%% iprop!
67 AV_RMSE (result, rvector) provides the metric of the root mean squared error
    .
68 %%% icalculate!
69 e_list = nncv.get('EVALUATOR_LIST');
70
71 value = cellfun(@(e) e.get('RMSE'), ...
72     e_list, 'UniformOutput', false);
73
74 if isempty(value)
75     value = [];
76 else
77     value = mean(cell2mat(value), 1);
78 end
79
80 %%% iprop!
81 AV_FEATURE_IMPORTANCE (result, cell) averages the feature importances across

```

```

        k folds. ⑥
82 %%% icalculate!
83 e_list = nncv.get('EVALUATOR_LIST');
84
85 all_fi = cellfun(@(e) cell2mat(e.get('FEATURE_IMPORTANCE')), ...
86     e_list, 'UniformOutput', false);
87
88 if isempty(cell2mat(all_fi))
89     value = {};
90 else
91     average_fi = zeros(size(all_fi{1}));
92     for i = 1:numel(all_fi)
93         % Add the current cell contents to the averageCell
94         average_fi = average_fi + all_fi{i};
95     end
96     average_fi = average_fi / numel(all_fi);
97     value = {average_fi};
98 end

```

⑥ calculate the average feature importance.

Code 4: NNRegressorMLP_CrossValidation element

tests. The tests section from the element generator `_NNRegressorMLP_CrossValidation.gen.m`. A test for creating example files should be prepared to test the properties of the data point. Furthermore, additional test should be prepared for validating the value of input and target for the data point.

```

1 %% itests!
2
3 %% itest!
4 %%% iname!
5 evaluate a regressor cross-validation with the example data
6 %%% icode!
7 % ensure the example data is generated
8 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
   REG CON XLS' filesep 'atlas.xlsx'])
9     test_NNDataPoint_CON_REG % create example files
10 end
11
12 % Load BrainAtlas
13 im_ba = ImporterBrainAtlasXLS( ...
14     'FILE', [fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data
   NN REG CON XLS' filesep 'atlas.xlsx'], ...
15     'WAITBAR', true ...
16 );
17
18 ba = im_ba.get('BA');
19
20 % Load Groups of SubjectCON
21 im_gr = ImporterGroupSubjectCON_XLS( ...
22     'DIRECTORY', [fileparts(which('NNDataPoint_CON_REG')) filesep 'Example
   data NN REG CON XLS' filesep 'CON_Group_XLS'], ...
23     'BA', ba, ...
24     'WAITBAR', true ...
25 );
26
27 gr = im_gr.get('GR');
28
29 % create a item list of NNDataPoint_CON_REG
30 it_list = cellfun(@(x) NNDataPoint_CON_REG( ...
31     'ID', x.get('ID'), ...
32     'SUB', x, ...
33     'TARGET_IDS', x.get('VOI_DICT').get('KEYS')), ...
34     gr.get('SUB_DICT').get('IT_LIST'), ...
35     'UniformOutput', false);
36
37 % create a NNDataPoint_CON_REG DICT
38 dp_list = IndexedDictionary(...
39     'IT_CLASS', 'NNDataPoint_CON_REG', ...
40     'IT_LIST', it_list ...
41 );
42
43 % create a NNData containing the NNDataPoint_CON_REG DICT
44 d = NNDataSet( ...
45     'DP_CLASS', 'NNDataPoint_CON_REG', ...
46     'DP_DICT', dp_list ...
47 );
48
49 kfold = 3;
50 nncv = NNRegressorMLP_CrossValidation('KFOLDS', kfold, 'D', d);

```

```
51
52 nn_list = nncv.get('NN_LIST');
53 assert(length(nn_list) == kfold, ... ①
54        [BRAPH2.STR ':NNRegressorMLP_CrossValidation:' BRAPH2.FAIL_TEST], ...
55        'NNRegressorMLP_CrossValidation does not calculate the neural network
56         list correctly.' ...
57        )
58 e_list = nncv.get('EVALUATOR_LIST');
59 assert(length(e_list) == kfold, ... ②
60        [BRAPH2.STR ':NNRegressorMLP_CrossValidation:' BRAPH2.FAIL_TEST], ...
61        'NNRegressorMLP_CrossValidation does not calculate the evaluator list
62         correctly.' ...
63        )
```

① and ② check whether the data, regressors, and evaluators are initialized according to the user-specified number of folds.

Implementation of the Cross Validation for Classifiers

We can now use `NNRegressorMLP_CrossValidation` as the basis to implement the `NNClassifierMLP_CrossValidation`. The parts of the code that are modified are highlighted. A cross validation for multi-layer perceptron classifier (`NNClassifierMLP_CrossValidation`) is a procedure designed for evaluating multi-layer perceptron classifiers using cross-validation.

Code 5: NNClassifierMLP_CrossValidation element header. The header section of the generator code for `_NNClassifierMLP_CrossValidation.gen.m` provides the general information about the `NNClassifierMLP_CrossValidation` element.

```

1 %% iheader!
2 NNClassifierMLP_CrossValidation < NNCrossValidation (nncv, neural network
   cross-validation) is a process for evaluating multi-layer perceptron
   classifiers using cross-validation.
3
4 %%% idescription!
5 A cross validation for multi-layer perceptron classifiers (
   NNClassifierMLP_CrossValidation) is a process that facilitates the
   evaluation of multi-layer perceptron classifiers using cross-validation
   .
6 It involves splitting a dataset into multiple subsets (folds), training the
   model on some folds while validating on others, and then repeating the
   process for all combinations of folds.
7 This helps in assessing the generalization performance of the model and
   detecting overfitting.
8
9 To train all the neural networks for all folds, use: nncv.get('TRAIN')

```

Code 6: NNClassifierMLP_CrossValidation element prop update. The `props_update` section of the generator code for `_NNClassifierMLP_CrossValidation.gen.m` updates the properties of the `NNClassifierMLP_CrossValidation` element. This defines the core properties of the data point.

```

1 %% iprops_update!
2
3 %%% iprop!
4 NAME (constant, string) is the name of the cross-validation.
5 %%% idefault!
6 'NNClassifierMLP_CrossValidation'
7
8 %%% iprop!
9 DESCRIPTION (constant, string) is the description of the cross-validation.
10 %%% idefault!
11 'A cross validation for multi-layer perceptron classifiers (
   NNClassifierMLP_CrossValidation) is a process that facilitates the
   evaluation of multi-layer perceptron classifiers using cross-validation
   . It involves splitting a dataset into multiple subsets (folds),
   training the model on some folds while validating on others, and then
   repeating the process for all combinations of folds. This helps in
   assessing the generalization performance of the model and detecting
   overfitting.'

```

```

12
13 %% iprop!
14 TEMPLATE (parameter, item) is the template of the cross-validation.
15 %%%% isettings!
16 'NNClassifierMLP_CrossValidation'
17
18 %% iprop!
19 ID (data, string) is a few-letter code for the cross-validation.
20 %%%% idefault!
21 'NNClassifierMLP_CrossValidation ID'
22
23 %% iprop!
24 LABEL (metadata, string) is an extended label of the cross-validation.
25 %%%% idefault!
26 'NNClassifierMLP_CrossValidation label'
27
28 %% iprop!
29 NOTES (metadata, string) are some specific notes about the cross-validation.
30 %%%% idefault!
31 'NNClassifierMLP_CrossValidation notes'
32
33 %% iprop!
34 NN_TEMPLATE (parameter, item) is the neural network template to set all
    neural network parameters.
35 %%%% isettings!
36 'NNClassifierMLP' ①
37
38 %% iprop!
39 NNEVALUATOR_TEMPLATE (parameter, item) is the neural network evaluator
    template to set all evaluator parameters.
40 %%%% isettings!
41 'NNClassifierMLP_Evaluator' ②
42
43 %% iprop!
44 NN_LIST (result, itemlist) contains the neural network models corresponding
    to k folds.
45 %%%% icalculate!
46 d_list = nncv.get('D_LIST');
47
48 if isempty(d_list)
49     value = {};
50 else
51     for i = 1:length(d_list)
52         d_training_set{i} = d_list;
53         d_training_set{i}(i) = []; % Exclude the i-th element
54         d_training_set{i} = NNDatasetCombine('D_LIST', d_training_set{i}).
            get('D');
55     end
56
57     d_training_set = d_training_set';
58
59     if ~isa(nncv.getr('NN_TEMPLATE'), 'NoValue')
60         nn_template = nncv.get('NN_TEMPLATE');
61     else
62         nn_template = NNClassifierMLP( ...
63             'EPOCHS', nncv.get('EPOCHS'), ...
64             'BATCH', nncv.get('BATCH'), ...
65             'SHUFFLE', nncv.get('SHUFFLE'), ...
66             'SOLVER', nncv.get('SOLVER'), ...
67             'VERBOSE', nncv.get('VERBOSE'), ...
68             'PLOT_TRAINING', nncv.get('PLOT_TRAINING'));

```

① and ② define the NN_TEMPLATE as NNClassifierMLP and the NNEVALUATOR_TEMPLATE as NNClassifierMLP_Evaluator.

```

69     end
70
71     value = cellfun(@(d) NNClassifierMLP( ...
72         'TEMPLATE', nn_template, 'D', d), ...
73         d_training_set, 'UniformOutput', false);
74 end
75
76 %%% iprop!
77 EVALUATOR_LIST (result, itemlist) contains the evaluators corresponding to k
78     folds.
79 %%% icalculate!
80 d_list = nncv.get('D_LIST');
81 nn_list = nncv.get('NN_LIST');
82
83 if ~isa(nncv.get('NNEVALUATOR_TEMPLATE'), 'NoValue')
84     nne_template = nncv.get('NNEVALUATOR_TEMPLATE');
85 else
86     nne_template = NNClassifierMLP_Evaluator( ...
87         'P', nncv.get('P'));
88 end
89
90 value = cellfun(@(d, nn) NNClassifierMLP_Evaluator('TEMPLATE', nne_template,
91     'D', d, 'NN', nn), ...
92     d_list, nn_list, 'UniformOutput', false);

```

Code 7: NNClassifierMLP_CrossValidation element props. The props section of generator code for `_NNClassifierMLP_CrossValidation.gen.m` defines the properties to be used in `NNClassifierMLP_CrossValidation`.

```

1  %% iprops!
2
3  %%% iprop!
4  P (parameter, scalar) is the permutation number.
5  %%% idefault!
6  1e+2
7  %%% icheck_prop!
8  check = value > 0 && value == round(value);
9
10 %%% iprop! ①
11 AV_AUC (result, rvector) provides the average value of the area under the
12     receiver operating characteristic curve across k folds.
13 %%% icalculate!
14 e_list = nncv.get('EVALUATOR_LIST');
15
16 aucs = cellfun(@(e) e.get('AUC'), ...
17     e_list, 'UniformOutput', false);
18
19 if isempty(aucs)
20     value = [];
21 else
22     value = mean(cell2mat(aucs), 1);
23 end
24
25 %%% iprop! ②
26 AV_MACRO_AUC (result, scalar) provides the metric of the average macro AUC
27     value across k folds.
28 %%% icalculate!
29 e_list = nncv.get('EVALUATOR_LIST');

```

① and ② calculate the average value of the area under the receiver operating characteristic curve across k folds.

```

28
29 macro_auc = cellfun(@(e) e.get('MACRO_AUC'), ...
30     e_list, 'UniformOutput', false);
31
32 if isempty(macro_auc)
33     value = 0;
34 else
35     value = mean(cell2mat(macro_auc), 1);
36 end
37
38 %% iprop! ③
39 C_MATRIX (result, matrix) provides the confusion matrix across k folds.
40 %%% icalculate!
41 e_list = nncv.get('EVALUATOR_LIST');
42
43 c_matrices = cellfun(@(e) e.get('C_MATRIX'), ...
44     e_list, 'UniformOutput', false);
45
46 combined_c_matrix = cellfun(@(x) double(x), c_matrices, 'UniformOutput',
47     false);
48 value = sum(cat(3, combined_c_matrix{:}), 3);
49
50 %% iprop!
51 AV_FEATURE_IMPORTANCE (result, cell) averages the feature importances across
52     k folds.
53 %%% icalculate!
54 e_list = nncv.get('EVALUATOR_LIST');
55
56 all_fi = cellfun(@(e) cell2mat(e.get('FEATURE_IMPORTANCE')), ...
57     e_list, 'UniformOutput', false);
58
59 if isempty(cell2mat(all_fi))
60     value = {};
61 else
62     average_fi = zeros(size(all_fi{1}));
63     for i = 1:numel(all_fi)
64         % Add the current cell contents to the averageCell
65         average_fi = average_fi + all_fi{i};
66     end
67     average_fi = average_fi / numel(all_fi);
68     value = {average_fi};
69 end

```

③ aggregates the confusion matrix across k folds.

Code 8: NNClassifierMLP_CrossValidation element

tests. The tests section from the element generator `_NNClassifierMLP_CrossValidation.gen.m`. A test for creating example files should be prepared to test the properties of the data point. Furthermore, additional test should be prepared for validating the value of input and target for the data point.

```

1 %% itests!
2
3 %% itest!
4 %%% iname!
5 evaluate a classifier cross-validation with the example data
6 %%% icode!
7
8 % ensure the example data is generated
9 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
   CLA CON XLS' filesep 'atlas.xlsx'])
10     test_NNDataPoint_CON_CLA % create example files
11 end
12
13 % Load BrainAtlas
14 im_ba = ImporterBrainAtlasXLS( ...
15     'FILE', [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data
   NN CLA CON XLS' filesep 'atlas.xlsx'], ...
16     'WAITBAR', true ...
17 );
18
19 ba = im_ba.get('BA');
20
21 % Load Groups of SubjectCON
22 im_gr1 = ImporterGroupSubjectCON_XLS( ...
23     'DIRECTORY', [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example
   data NN CLA CON XLS' filesep 'CON_Group_1_XLS'], ...
24     'BA', ba, ...
25     'WAITBAR', true ...
26 );
27
28 gr1 = im_gr1.get('GR');
29
30 im_gr2 = ImporterGroupSubjectCON_XLS( ...
31     'DIRECTORY', [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example
   data NN CLA CON XLS' filesep 'CON_Group_2_XLS'], ...
32     'BA', ba, ...
33     'WAITBAR', true ...
34 );
35
36 gr2 = im_gr2.get('GR');
37
38 % create item lists of NNDataPoint_CON_CLA
39 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
40 it_list1 = cellfun(@(x) NNDataPoint_CON_CLA( ...
41     'ID', x.get('ID'), ...
42     'SUB', x, ...
43     'TARGET_IDS', {group_folder_name}), ...
44     gr1.get('SUB_DICT').get('IT_LIST'), ...
45     'UniformOutput', false);
46
47 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
48 it_list2 = cellfun(@(x) NNDataPoint_CON_CLA( ...
49     'ID', x.get('ID'), ...

```

```

50     'SUB', x, ...
51     'TARGET_IDS', {group_folder_name}), ...
52     gr2.get('SUB_DICT').get('IT_LIST'), ...
53     'UniformOutput', false);
54
55 % create NNDataPoint_CON_CLA DICT items
56 dp_list1 = IndexedDictionary(...
57     'IT_CLASS', 'NNDataPoint_CON_CLA', ...
58     'IT_LIST', it_list1 ...
59 );
60
61 dp_list2 = IndexedDictionary(...
62     'IT_CLASS', 'NNDataPoint_CON_CLA', ...
63     'IT_LIST', it_list2 ...
64 );
65
66 % create a NNDataset containing the NNDataPoint_CON_CLA DICT
67 d1 = NNDataset( ...
68     'DP_CLASS', 'NNDataPoint_CON_CLA', ...
69     'DP_DICT', dp_list1 ...
70 );
71
72 d2 = NNDataset( ...
73     'DP_CLASS', 'NNDataPoint_CON_CLA', ...
74     'DP_DICT', dp_list2 ...
75 );
76
77 % combine the two datasets
78 d = NNDatasetCombine('D_LIST', {d1, d2}).get('D');
79
80 kfolgs = 7;
81 nncv = NNClassifierMLP_CrossValidation('KFOLDS', kfolgs, 'D', d);
82
83 nn_list = nncv.get('NN_LIST');
84 assert(length(nn_list) == kfolgs, ... ①
85     [BRAPH2.STR 'NNClassifierMLP_CrossValidation:' BRAPH2.FAIL_TEST], ...
86     'NNClassifierMLP_CrossValidation does not calculate the neural network
87     list correctly.' ...
88 );
89 e_list = nncv.get('EVALUATOR_LIST');
90 assert(length(e_list) == kfolgs, ... ②
91     [BRAPH2.STR 'NNClassifierMLP_CrossValidation:' BRAPH2.FAIL_TEST], ...
92     'NNClassifierMLP_CrossValidation does not calculate the evaluator list
93     correctly.' ...
94 );

```

① and ② check whether the data, classifiers, and evaluators are initialized according to the user-specified number of folds.