

# *Implement a New Neural Network Data Point*

*The BRAPH 2 Developers*

*August 14, 2024*

This is the developer tutorial for implementing a new neural network data point. In this Tutorial, you will learn how to create the generator file `*.gen.m` for a new neural network data point, which can then be compiled by `braph2genesis`. All kinds of neural network data point are (direct or indirect) extensions of the base element `NNDataPoint`. Here, you will use as examples the neural network data point `NNDataPoint_CON_REG` (connectivity data for regression), `NNDataPoint_CON_CLA` (connectivity data for classification), `NNDataPoint_Graph_REG` (adjacency matrix for regression), `NNDataPoint_Graph_CLA` (adjacency matrix for classification), `NNDataPoint_Measure_REG` (graph measure for regression), and `NNDataPoint_Measure_CLA` (graph measure for classification).

## *Contents*

<i>Implementation of a data point with connectivity data</i>	2
<i>Connectivity data point for regression (NNDataPoint_CON_REG)</i>	2
<i>Connectivity data point for classification (NNDataPoint_CON_CLA)</i>	8
<i>Implementation of a data point with a graph</i>	15
<i>Graph data point for regression (NNDataPoint_Graph_REG)</i>	15
<i>Graph data point for classification (NNDataPoint_Graph_CLA)</i>	21
<i>Implementation of a data point with graph measures</i>	27
<i>Graph measure data point for regression (NNDataPoint_Measure_REG)</i>	27
<i>Graph measure data point for classification (NNDataPoint_Measure_CLA)</i>	32

## Implementation of a data point with connectivity data

### Connectivity data point for regression (NNDataPoint\_CON\_REG)

You will start by implementing in detail NNDataPoint\_CON\_REG, which is a direct extension of NNDataPoint. A NNDataPoint\_CON\_REG contains the input and target for neural network analysis with a subject with connectivity data (SubjectCON), where the input is the subject's connectivity data and the target are the subject's variables of interest.

**Code 1: NNDataPoint\_CON\_REG element header.** The header section of the generator code in \_NNDataPoint\_CON\_REG.gen.m provides the general information about the NNDataPoint\_CON\_REG element.

---

```

1 %% iheader!
2 NNDataPoint_CON_REG < NNDataPoint (dp, connectivity regression data point)
   is a data point for regression with connectivity data.
3
4
5 %% idescription!
6 A data point for regression with connectivity data (NNDataPoint_CON_REG)
7 contains the input and target for neural network analysis with a subject
   with connectivity data (SubjectCON).
8 The input is the connectivity data of the subject.
9 The target is obtained from the variables of interest of the subject.
10
11 %% ibuild!
12 1

```

---

① defines NNDataPoint\_CON\_REG as a subclass of NNDataPoint. The moniker will be dp.

**Code 2: NNDataPoint\_CON\_REG element props update.** The props\_update section of the generator code in \_NNDataPoint\_CON\_REG.gen.m updates the properties of the NNDataPoint element. This defines the core properties of the data point.

---

```

1 %% iprops_update!
2
3 %% iprop!
4 NAME (constant, string) is the name of a data point for regression with
   connectivity data.
5 %%% idefault!
6 'NNDataPoint_CON_REG'
7
8 %% iprop!
9 DESCRIPTION (constant, string) is the description of a data point for
   regression with connectivity data.
10 %%% idefault!
11 'A data point for regression with connectivity data (NNDataPoint_CON_REG)
   contains the input and target for neural network analysis with a
   subject with connectivity data (SubjectCON). The input is the
   connectivity data of the subject. The target is obtained from the
   variables of interest of the subject.'
12
13 %%% iprop!
14 TEMPLATE (parameter, item) is the template of a data point for regression
   with connectivity data.

```

---

```

15 %%% isettings!
16 'NNDataPoint_CON_REG'
17
18 %%% iprop!
19 ID (data, string) is a few-letter code for a data point for regression with
    connectivity data.
20 %%% idefault!
21 'NNDataPoint_CON_REG ID'
22
23 %%% iprop!
24 LABEL (metadata, string) is an extended label of a data point for regression
    with connectivity data.
25 %%% idefault!
26 'NNDataPoint_CON_REG label'
27
28 %%% iprop!
29 NOTES (metadata, string) are some specific notes about a data point for
    regression with connectivity data.
30 %%% idefault!
31 'NNDataPoint_CON_REG notes'
32
33 %%% iprop! ①
34 INPUT (result, cell) is the input value for this data point.
35 %%% icalculate!
36 value = {dp.get('SUB').get('CON')}; ②
37
38 %%% iprop! ③
39 TARGET (result, cell) is the target value for this data point.
40 %%% icalculate!
41 value = cellfun(@(x) dp.get('SUB').get('VOI_DICT').get('IT', x).get('V'), dp
    .get('TARGET_IDS'), 'UniformOutput', false); ④

```

① The property INPUT is the input value for this data point, which is obtained directly from the connectivity data of Subject\_CON ②.

③ The property TARGET is the target value for this data point, which is obtained directly from the variables of interest ④.

---

Code 3: NNDataPoint\_CON\_REG element props. The props section of generator code in NNDataPoint\_CON\_REG.gen.m defines the properties to be used in NNDataPoint\_CON\_REG.

---

```

1 %%% iprops!
2
3 %%% iprop! ①
4 SUB (data, item) is a subject with connectivity data.
5 %%% isettings!
6 'SubjectCON'
7
8 %%% iprop! ②
9 TARGET_IDS (parameter, stringlist) is a list of variable-of-interest IDs to
    be used as regression targets.

```

① The property SUB is a subject with connectivity data (Subject\_CON), which is used to calculate the properties INPUT and TARGET.

② The property TARGET\_IDS defines the IDs of target, where the target IDs should be from the subject's variable-of-interest IDs.

Code 4: **NNDataPoint\_CON\_REG element tests**. The tests section from the element generator `_NNDataPoint_CON_REG.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 %%% iexcluded_props! ①
4 [NNDataPoint_CON_REG.SUB]
5
6 %%% itest!
7 %%% iname!
8 Create example files for regression ②
9 %%% icode!
10 data_dir = [fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
    REG CON XLS'];
11 if ~isdir(data_dir)
12     mkdir(data_dir);
13
14 % Brain Atlas ③
15 im_ba = ImporterBrainAtlasXLS('FILE', 'desikan.atlas.xlsx');
16 ba = im_ba.get('BA');
17 ex_ba = ExporterBrainAtlasXLS( ...
18     'BA', ba, ...
19     'FILE', [data_dir filesep() 'atlas.xlsx'] ...
20 );
21 ex_ba.get('SAVE')
22 N = ba.get('BR_DICT').get('LENGTH');
23
24 % saves RNG
25 rng_settings_ = rng(); rng('default')
26
27 sex_options = {'Female' 'Male'};
28
29 % Group ④
30 K = 2; % degree (mean node degree is 2K)
31 beta = 0.3; % Rewiring probability
32 gr_name = 'CON_Group_XLS';
33 gr_dir = [data_dir filesep() gr_name];
34 mkdir(gr_dir);
35 vois = [
36     {'Subject ID'} {'Age'} {'Sex'}
37     {} {} cell2str(sex_options)}
38 ];
39 for i = 1:100 % subject number
40     sub_id = ['SubjectCON_' num2str(i)];
41     % create WS graphs with random beta
42     beta(i) = rand(1); ⑤
43     h = WattsStrogatz(N, K, beta(i)); % create WS graph ⑥
44
45     A = full(adjacency(h)); A(1:length(A)+1:numel(A)) = 0; % extract the
    adjacency matrix
46     r = 0 + (0.5 - 0) * rand(size(A)); diffA = A - r; A(A ~= 0) = diffA(
    A ~= 0); % make the adjacency matrix weighted
47     A = max(A, transpose(A)); % make the adjacency matrix symmetric
48
49     writetable(array2table(A), [gr_dir filesep() sub_id '.xlsx'], '

```

① List of properties that are excluded from testing.

② creates the example connectivity data files for regression analysis.

③ creates and exports the brain atlas file to the example directory.

④ creates one group of subjects with specified degree and rewiring probability configurations.

⑤ generates random rewiring probability settings for each subject.

⑥ utilizes the provided degree and rewiring probability settings to generate corresponding Watts-Strogatz model graphs using the function `WattsStrogatz` ⑩.

```

    WriteVariableNames', false) ⑦
50
51     % variables of interest
52     age_upperBound = 80;
53     age_lowerBound = 50;
54     age = age_lowerBound + beta(i)*(age_upperBound - age_lowerBound);
    ⑧
55     vois = [vois; {sub_id, age, sex_options(randi(2))}];
56 end
57 writetable(table(vois), [data_dir filesep() gr_name '.vois.xlsx'], '
    WriteVariableNames', false) ⑨
58
59 % reset RNG
60 rng(rng_settings_)
61 end
62 %%% itest_functions!
63 function h = WattsStrogatz(N, K, beta) ⑩
64 % H = WattsStrogatz(N,K,beta) returns a Watts-Strogatz model graph with N
65 % nodes, N*K edges, mean node degree 2*K, and rewiring probability beta.
66 %
67 % beta = 0 is a ring lattice, and beta = 1 is a random graph.
68
69 % Connect each node to its K next and previous neighbors. This constructs
70 % indices for a ring lattice.
71 s = repelem((1:N)', 1, K);
72 t = s + repmat(1:K, N, 1);
73 t = mod(t - 1, N) + 1;
74
75 % Rewire the target node of each edge with probability beta
76 for source = 1:N
77     switchEdge = rand(K, 1) < beta;
78
79     newTargets = rand(N, 1);
80     newTargets(source) = 0;
81     newTargets(s(t == source)) = 0;
82     newTargets(t(source, ~switchEdge)) = 0;
83
84     [~, ind] = sort(newTargets, 'descend');
85     t(source, switchEdge) = ind(1:nz(switchEdge));
86 end
87
88 h = graph(s,t);
89 end
90
91 %%% itest!
92 %%%% iname! ⑪
93 Create a NNDataset containing NNDatapoint_CON_REG with simulated data
94 %%%% icode!
95 % Load BrainAtlas
96 im_ba = ImporterBrainAtlasXLS( ...
97     'FILE', [fileparts(which('NNDatapoint_CON_REG')) filesep() 'Example data
    NN REG CON XLS' filesep() 'atlas.xlsx'], ...
98     'WAITBAR', true ...
99     );
100
101 ba = im_ba.get('BA');
102
103 % Load Group of SubjectCON
104 im_gr = ImporterGroupSubjectCON_XLS( ...

```

⑦ exports the adjacency matrix of the graph to an Excel file.

⑧ associates the age value with each individual rewiring probability setting.

⑨ exports the variables of interest to an Excel file.

⑪ validates the data point by using assertions to confirm that the input and target calculated values match the connectivity data and the variables of interest in the example files.

```

105     'DIRECTORY', [fileparts(which('NNDataPoint_CON_REG')) filesep 'Example
        data NN REG CON XLS' filesep 'CON_Group_XLS'], ...
106     'BA', ba, ...
107     'WAITBAR', true ...
108     );
109
110 gr = im_gr.get('GR');
111
112 % create an item list of NNDataPoint_CON_REG (12)
113 it_list = cellfun(@(x) NNDataPoint_CON_REG( ...
114     'ID', x.get('ID'), ...
115     'SUB', x, ...
116     'TARGET_IDS', x.get('VOI_DICT').get('KEYS')), ...
117     gr.get('SUB_DICT').get('IT_LIST'), ...
118     'UniformOutput', false);
119
120 % create a NNDataPoint_CON_REG DICT (13)
121 dp_list = IndexedDictionary(...
122     'IT_CLASS', 'NNDataPoint_CON_REG', ...
123     'IT_LIST', it_list ...
124     );
125
126 % create a NNDataset containing the NNDataPoint_CON_REG DICT (14)
127 d = NNDataset( ...
128     'DP_CLASS', 'NNDataPoint_CON_REG', ...
129     'DP_DICT', dp_list ...
130     );
131
132 % Check whether the number of inputs matches (14)
133 assert(length(d.get('INPUTS')) == gr.get('SUB_DICT').get('LENGTH'), ...
134     [BRAPH2.STR ':NNDataPoint_CON_REG:' BRAPH2.FAIL_TEST], ...
135     'NNDataPoint_CON_REG does not construct the dataset correctly. The
        number of the inputs should be the same as the number of imported
        subjects.' ...
136     )
137
138 % Check whether the number of targets matches (15)
139 assert(length(d.get('TARGETS')) == gr.get('SUB_DICT').get('LENGTH'), ...
140     [BRAPH2.STR ':NNDataPoint_CON_REG:' BRAPH2.FAIL_TEST], ...
141     'NNDataPoint_CON_REG does not construct the dataset correctly. The
        number of the targets should be the same as the number of imported
        subjects.' ...
142     )
143
144 % Check whether the content of input for a single data point matches (16)
145 for index = 1:l:gr.get('SUB_DICT').get('LENGTH')
146     individual_input = d.get('DP_DICT').get('IT', index).get('INPUT');
147     known_input = {gr.get('SUB_DICT').get('IT', index).get('CON')};
148
149     assert(isequal(individual_input, known_input), ...
150         [BRAPH2.STR ':NNDataPoint_CON_REG:' BRAPH2.FAIL_TEST], ...
151         'NNDataPoint_CON_REG does not construct the dataset correctly. The
            input value is not derived correctly.' ...
152         )
153 end
154
155 %% itest!

```

(12)–(14) create an item list for the data points, subsequently generates the data point dictionary using the list, and then constructs the neural network dataset containing these data points.

(14) tests the number of inputs from the dataset matches the number of subjects in the group.

(15) tests the number of targets from the dataset matches the number of subjects in the group.

(16) tests the value of each input from the data point matches the subject's connectivity data.

```
156 %%%% iname! ①7
157 Example training-test regression
158 %%%% icode!
159 % ensure the example data is generated
160 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
    REG CON XLS' filesep 'atlas.xlsx'])
161     test_NNDataPoint_CON_REG % create example files
162 end
163
164 example_NN_CON_REG
```

---

①7 executes the corresponding example scripts to ensure the functionalities.

### *Connectivity data point for classification (NNDataPoint\_CON\_CLA)*

You can now use NNDataPoint\_CON\_REG as the basis to implement the NNDataPoint\_CON\_CLA, which can be used for classification tasks. The parts of the code that are modified are highlighted.

**Code 5: NNDataPoint\_CON\_CLA element header.** The header section of the generator code in `_NNDataPoint_CON_CLA.gen.m` provides the general information about the NNDataPoint\_CON\_CLA element. ← [Code 1](#)

---

```

1 %% iheader!
2 NNDataPoint_CON_CLA < NNDataPoint (dp, connectivity classification data
    point) is a data point for classification with connectivity data.
3
4 %% idescription!
5 A data point for classification with connectivity data (NNDataPoint_CON_CLA)
6 contains the input and target for neural network analysis with a subject
    with connectivity data (SubjectCON).
7 The input is the connectivity data of the subject.
8 The target is obtained from the variables of interest of the subject.
9
10 %% ibuild!
11 1

```

---

**Code 6: NNDataPoint\_CON\_CLA element props update.** The props\_update section of the generator code in `_NNDataPoint_CON_CLA.gen.m` updates the properties of the NNDataPoint\_CON\_CLA element. This defines the core properties of the data point. ← [Code 2](#)

---

```

1 %% iprops_update!
2
3 %% iprop!
4 NAME (constant, string) is the name of a data point for classification with
    connectivity data.
5 %%% idefault!
6 'NNDataPoint_CON_CLA'
7
8 %% iprop!
9 DESCRIPTION (constant, string) is the description of a data point for
    classification with connectivity data.
10 %%% idefault!
11 'A data point for classification with connectivity data (NNDataPoint_CON_CLA
    ) contains the input and target for neural network analysis with a
    subject with connectivity data (SubjectCON). The input is the
    connectivity data of the subject. The target is obtained from the
    variables of interest of the subject.'
12
13 %% iprop!
14 TEMPLATE (parameter, item) is the template of a data point for
    classification with connectivity data.
15 %%% isettings!
16 'NNDataPoint_CON_CLA'
17
18 %% iprop!
19 ID (data, string) is a few-letter code for a data point for classification
    with connectivity data.

```

---



```

20 %%%% idefault!
21 'NNDataPoint_CON_CLA ID'
22
23 %%% iprop!
24 LABEL (metadata, string) is an extended label of a data point for
    classification with connectivity data.
25 %%%% idefault!
26 'NNDataPoint_CON_CLA label'
27
28 %%% iprop!
29 NOTES (metadata, string) are some specific notes about a data point for
    classification with connectivity data.
30 %%%% idefault!
31 'NNDataPoint_CON_CLA notes'
32
33 %%% iprop!
34 INPUT (result, cell) is the input value for this data point.
35 %%%% icalculate!
36 value = {dp.get('SUB').get('CON')};
37
38 %%% iprop! ①
39 TARGET (result, cell) is the target values for this data point. ②
40 %%%% icalculate!
41 value = {dp.get('TARGET_CLASS')}; ③

```

---

Code 7: **NNDataPoint\_CON\_CLA element props.** The props section of generator code in `_NNDataPoint_CON_CLA.gen.m` defines the properties to be used in `NNDataPoint_CON_CLA`.

---

① defines the target value using the data point's label in the form of a string, e.g., 'Group1', contained in a cell array and representing the class of the data point.

② changed stringlist to cell.

③ changed prop name.

```

1 %%% iprops!
2
3 %%% iprop!
4 SUB (data, item) is a subject with connectivity data.
5 %%%% issettings!
6 'SubjectCON'
7
8 %%% iprop! ④
9 TARGET_CLASS (parameter, string) is the name of the class of the data point.

```

④ changed. revise also the tests.

---

Code 8: **NNDataPoint\_CON\_CLA element tests.** The tests section from the element generator `_NNDataPoint_CON_CLA.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 %%% iexcluded_props!
4 [NNDataPoint_CON_CLA.SUB]
5
6 %%% itest!
7 %%%% iname!
8 Create example files
9 %%%% icode!
10 data_dir = [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
    CLA CON XLS'];
11 if ~isdir(data_dir)

```

```

12 mkdir(data_dir);
13
14 ...
15
16 % Group 1 ①
17 K1 = 2; % degree (mean node degree is 2K) - group 1
18 beta1 = 0.3; % Rewiring probability - group 1
19 gr1_name = 'CON_Group_1_XLS';
20 gr1_dir = [data_dir filesep() gr1_name];
21 mkdir(gr1_dir);
22 vois1 = [
23     {'Subject ID'} {'Age'} {'Sex'}}
24     {''} {''} cell2str(sex_options)}
25     ];
26 for i = 1:1:50 % subject number
27     sub_id = ['SubjectCON_' num2str(i)];
28
29     h1 = WattsStrogatz(N, K1, beta1); % create two WS graph
30     figure(1) % Plot the two graphs to double-check
31     % plot(h1, 'NodeColor',[1 0 0], 'EdgeColor',[0 0 0], 'EdgeAlpha
32     % title(['Group 1: Graph with $N = $ ' num2str(N_nodes) ...
33     %       ' nodes, $K = $ ' num2str(K1) ', and $\beta = $ ' num2str(
34     %       'Interpreter','latex')
35     % axis equal
36
37     A1 = full(adjacency(h1)); A1(1:length(A1)+1:numel(A1)) = 0; %
38     % extract the adjacency matrix
39     r = 0 + (0.5 - 0)*rand(size(A1)); diffA = A1 - r; A1(A1 ~= 0) =
40     % diffA(A1 ~= 0); % make the adjacency matrix weighted
41     A1 = max(A1, transpose(A1)); % make the adjacency matrix symmetric
42
43     writetable(array2table(A1), [gr1_dir filesep() sub_id '.xlsx'], '
44     % WriteVariableNames', false)
45
46     % variables of interest
47     vois1 = [vois1; {sub_id, randi(90), sex_options(randi(2))}];
48 end
49 writetable(table(vois1), [data_dir filesep() gr1_name '.vois.xlsx'], '
50 % WriteVariableNames', false)
51
52 % Group 2 ②
53 K2 = 2; % degree (mean node degree is 2K) - group 2
54 beta2 = 0.85; % Rewiring probability - group 2 ③
55 gr2_name = 'CON_Group_2_XLS';
56 gr2_dir = [data_dir filesep() gr2_name];
57 mkdir(gr2_dir);
58 vois2 = [
59     {'Subject ID'} {'Age'} {'Sex'}}
60     {''} {''} cell2str(sex_options)}
61     ];
62 for i = 51:1:100
63     sub_id = ['SubjectCON_' num2str(i)];
64
65     h2 = WattsStrogatz(N, K2, beta2);
66     figure(2)
67     % plot(h2, 'NodeColor',[1 0 0], 'EdgeColor',[0 0 0], 'EdgeAlpha
68     % title(['Group 2: Graph with $N = $ ' num2str(N_nodes) ...

```

① creates the first group of simulated data.

② and ③ create the second group of simulated data with different rewiring probability parameter.

```

65     %     ' nodes, $K = $ ' num2str(K2) ', and $\beta = $ ' num2str(
beta2)], ...
66     %     'Interpreter','latex')
67     % axis equal
68
69     A2 = full(adjacency(h2)); A2(1:length(A2)+1:numel(A2)) = 0;
70     r = 0 + (0.5 - 0)*rand(size(A2)); diffA = A2 - r; A2(A2 ~= 0) =
diffA(A2 ~= 0);
71     A2 = max(A2, transpose(A2));
72
73     writetable(array2table(A2), [gr2_dir filesep() sub_id '.xlsx'], '
WriteVariableNames', false)
74
75     % variables of interest
76     vois2 = [vois2; {sub_id, randi(90), sex_options(randi(2)))}];
77 end
78 writetable(table(vois2), [data_dir filesep() gr2_name '.vois.xlsx'], '
WriteVariableNames', false)
79
80 % Group 3 ④
81 K3 = 2; % degree (mean node degree is 2K) - group 3
82 beta3 = 0.55; % Rewiring probability - group 3 ⑤
83 gr3_name = 'CON_Group_3_XLS';
84 gr3_dir = [data_dir filesep() gr3_name];
85 mkdir(gr3_dir);
86 vois3 = [
87     {'Subject ID'} {'Age'} {'Sex'}
88     {} {} cell2str(sex_options)
89     ];
90 for i = 101:1:150
91     sub_id = ['SubjectCON_' num2str(i)];
92
93     h3 = WattsStrogatz(N, K3, beta3);
94     % figure(2)
95     % plot(h2, 'NodeColor',[1 0 0], 'EdgeColor',[0 0 0], 'EdgeAlpha
',0.1, 'Layout','circle');
96     % title(['Group 2: Graph with $N = $ ' num2str(N_nodes) ...
97     %     ' nodes, $K = $ ' num2str(K2) ', and $\beta = $ ' num2str(
beta2)], ...
98     %     'Interpreter','latex')
99     % axis equal
100
101     A3 = full(adjacency(h3)); A3(1:length(A3)+1:numel(A3)) = 0;
102     r = 0 + (0.5 - 0)*rand(size(A3)); diffA = A3 - r; A3(A3 ~= 0) =
diffA(A3 ~= 0);
103     A3 = max(A3, transpose(A3));
104
105     writetable(array2table(A3), [gr3_dir filesep() sub_id '.xlsx'], '
WriteVariableNames', false)
106
107     % variables of interest
108     vois3 = [vois3; {sub_id, randi(90), sex_options(randi(2)))}];
109 end
110 writetable(table(vois3), [data_dir filesep() gr3_name '.vois.xlsx'], '
WriteVariableNames', false)
111
112 % reset RNG
113 rng(rng_settings_)
114 end
115
116 %%% itest_functions!

```

④ and ⑤ create the third group of simulated data with different rewiring probability parameter.

```

117 function h = WattsStrogatz(N,K,beta)
118 ...
119
120 %% itest!
121 %%% iname!
122 Create a NNDataSet containing NNDataPoint_CON_CLA with simulated data
123 %%% icode!
124 % Load BrainAtlas
125 im_ba = ImporterBrainAtlasXLS( ...
126     'FILE', [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data
127     NN CLA CON XLS' filesep 'atlas.xlsx'], ...
128     'WAITBAR', true ...
129 );
130 ba = im_ba.get('BA');
131
132 % Load Groups of SubjectCON ⑥
133 im_gr1 = ImporterGroupSubjectCON_XLS( ...
134     'DIRECTORY', [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example
135     data NN CLA CON XLS' filesep 'CON_Group_1_XLS'], ...
136     'BA', ba, ...
137     'WAITBAR', true ...
138 );
139 gr1 = im_gr1.get('GR');
140
141 im_gr2 = ImporterGroupSubjectCON_XLS( ...
142     'DIRECTORY', [fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example
143     data NN CLA CON XLS' filesep 'CON_Group_2_XLS'], ...
144     'BA', ba, ...
145     'WAITBAR', true ...
146 );
147 gr2 = im_gr2.get('GR');
148
149 % create item lists of NNDataPoint_CON_CLA ⑦
150 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
151 it_list1 = cellfun(@(x) NNDataPoint_CON_CLA( ...
152     'ID', x.get('ID'), ...
153     'SUB', x, ...
154     'TARGET_CLASS', group_folder_name), ... ⑧
155     gr1.get('SUB_DICT').get('IT_LIST'), ...
156     'UniformOutput', false);
157
158 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
159 it_list2 = cellfun(@(x) NNDataPoint_CON_CLA( ...
160     'ID', x.get('ID'), ...
161     'SUB', x, ...
162     'TARGET_CLASS', group_folder_name), ... ⑨
163     gr2.get('SUB_DICT').get('IT_LIST'), ...
164     'UniformOutput', false);
165
166 % create NNDataPoint_CON_CLA DICT items
167 dp_list1 = IndexedDictionary(...
168     'IT_CLASS', 'NNDataPoint_CON_CLA', ...
169     'IT_LIST', it_list1 ...
170 );
171
172 dp_list2 = IndexedDictionary(...
173     'IT_CLASS', 'NNDataPoint_CON_CLA', ...

```

⑥ imports two groups of simulated data.

⑦ creates two datasets for the two groups.

⑧ changed.

⑨ changed.

```

174         'IT_LIST', it_list2 ...
175     );
176
177 % create a NNDataSet containing the NNDataPoint_CON_CLA DICT
178 d1 = NNDataSet( ...
179     'DP_CLASS', 'NNDataPoint_CON_CLA', ...
180     'DP_DICT', dp_list1 ...
181 );
182
183 d2 = NNDataSet( ...
184     'DP_CLASS', 'NNDataPoint_CON_CLA', ...
185     'DP_DICT', dp_list2 ...
186 );
187
188 % Check whether the number of inputs matches ⑧
189 assert(length(d1.get('INPUTS')) == gr1.get('SUB_DICT').get('LENGTH'), ...
190     [BRAPH2.STR ':NNDataPoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
191     'NNDataPoint_CON_CLA does not construct the dataset correctly. The
192     number of the inputs should be the same as the number of imported
193     subjects of group 1.' ...
194 );
195
196 assert(length(d2.get('INPUTS')) == gr2.get('SUB_DICT').get('LENGTH'), ...
197     [BRAPH2.STR ':NNDataPoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
198     'NNDataPoint_CON_CLA does not construct the dataset correctly. The
199     number of the inputs should be the same as the number of imported
200     subjects of group 2.' ...
201 );
202
203 % Check whether the number of targets matches ⑨
204 assert(length(d1.get('TARGETS')) == gr1.get('SUB_DICT').get('LENGTH'), ...
205     [BRAPH2.STR ':NNDataPoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
206     'NNDataPoint_CON_CLA does not construct the dataset correctly. The
207     number of the targets should be the same as the number of imported
208     subjects of group 1.' ...
209 );
210
211 assert(length(d2.get('TARGETS')) == gr2.get('SUB_DICT').get('LENGTH'), ...
212     [BRAPH2.STR ':NNDataPoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
213     'NNDataPoint_CON_CLA does not construct the dataset correctly. The
214     number of the targets should be the same as the number of imported
215     subjects of group 2.' ...
216 );
217
218 % Check whether the content of input for a single data point matches ⑩
219 for index = 1:1:gr1.get('SUB_DICT').get('LENGTH')
220     individual_input = d1.get('DP_DICT').get('IT', index).get('INPUT');
221     known_input = {gr1.get('SUB_DICT').get('IT', index).get('CON')};
222
223     assert(isequal(individual_input, known_input), ...
224         [BRAPH2.STR ':NNDataPoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
225         'NNDataPoint_CON_CLA does not construct the dataset correctly. The
226         input value is not derived correctly.' ...
227     );
228 end
229
230 for index = 1:1:gr2.get('SUB_DICT').get('LENGTH')
231     individual_input = d2.get('DP_DICT').get('IT', index).get('INPUT');
232     known_input = {gr2.get('SUB_DICT').get('IT', index).get('CON')};
233
234     assert(isequal(individual_input, known_input), ...
235         [BRAPH2.STR ':NNDataPoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
236         'NNDataPoint_CON_CLA does not construct the dataset correctly. The
237         input value is not derived correctly.' ...
238     );
239 end

```

⑧ tests the number of inputs from the dataset matches the number of subjects in the group.

⑨ tests the number of targets from the dataset matches the number of subjects in the group.

⑩ tests the value of each input from the data point matches the subject's connectivity data.

```

225     assert(isequal(individual_input, known_input), ...
226           [BRAPH2.STR ':NNDatapoint_CON_CLA:' BRAPH2.FAIL_TEST], ...
227           'NNDatapoint_CON_CLA does not construct the dataset correctly. The
           input value is not derived correctly.' ...
228           )
229 end
230
231 %%% itest!
232 %%% iname!
233 Example training-test classification (11)
234 %%% icode!
235 % ensure the example data is generated
236 if ~isfile([fileparts(which('NNDatapoint_CON_CLA')) filesep 'Example data NN
           CLA CON XLS' filesep 'atlas.xlsx'])
237     test_NNDatapoint_CON_CLA % create example files
238 end
239
240 example_NN_CON_CLA (0)

```

---

(11) executes the corresponding example scripts to ensure the functionalities.

(0) revise next ones to take into account the changes.

## *Implementation of a data point with a graph*

### *Graph data point for regression (NNDataPoint\_Graph\_REG)*

Now you will implement `NNDataPoint_Graph_REG` based on previous codes `NNDataPoint_CON_REG`. This neural network data point with graphs utilizes the adjacency matrix extracted from the derived graph of the subject. The modified parts of the code are highlighted.

**Code 9: NNDataPoint\_Graph\_REG element header.** The header section of the generator code in `_NNDataPoint_Graph_REG.gen.m` provides the general information about the `NNDataPoint_Graph_REG` element.

---

```

1 %% iheader!
2 NNDataPoint_Graph_REG < NNDataPoint (dp, measure regression) is
  a data point for regression with a graph.
3
4 %% idescription!
5 A data point for regression with a graph (NNDataPoint_Graph_REG)
6 contains both input and target for neural network analysis.
7 The input is the value of the adjacency matrix extracted from the derived
  graph of the subject.
8 The target is obtained from the variables of interest of the subject.
9
10 %% ibuild!
11 1

```

---

**Code 10: NNDataPoint\_Graph\_REG element props update.** The `props_update` section of the generator code in `_NNDataPoint_Graph_REG.gen.m` updates the properties of the `NNDataPoint_Graph_REG` element. This defines the core properties of the data point.

---

```

1 %% iprops_update!
2
3 %% iprop!
4 NAME (constant, string) is the name of a data point for regression with a
  graph.
5 %%% idefault!
6 'NNDataPoint_Graph_REG'
7
8 %%% iprop!
9 DESCRIPTION (constant, string) is the description of a data point for
  regression with a graph.
10 %%% idefault!
11 'A data point for regression with a graph (NNDataPoint_Graph_REG) contains
  both input and target for neural network analysis. The input is the
  value of the adjacency matrix extracted from the derived graph of the
  subject. The target is obtained from the variables of interest of the
  subject.'
12
13 %%% iprop!
14 TEMPLATE (parameter, item) is the template of a data point for regression
  with a graph.
15 %%% isettings!

```

---

```

16 'NNDataPoint_Graph_REG'
17
18 %% iprop!
19 ID (data, string) is a few-letter code for a data point for regression with
    a graph.
20 %%%% idefault!
21 'NNDataPoint_Graph_REG ID'
22
23 %% iprop!
24 LABEL (metadata, string) is an extended label of a data point for regression
    with a graph.
25 %%%% idefault!
26 'NNDataPoint_Graph_REG label'
27
28 %% iprop!
29 NOTES (metadata, string) are some specific notes about a data point for
    regression with a graph.
30 %%%% idefault!
31 'NNDataPoint_Graph_REG notes'
32
33 %% iprop!
34 INPUT (result, cell) is the input value for this data point.
35 %%%% icalculate!
36 value = dp.get('G').get('A'); ①
37
38 %% iprop!
39 TARGET (result, cell) is the target value for this data point.
40 %%%% icalculate!
41 value = cellfun(@(x) dp.get('SUB').get('VOI_DICT').get('IT', x).get('V'), dp
    .get('TARGET_IDS'), 'UniformOutput', false);

```

① extracts the adjacency matrix from a Graph element as the input for this data point. Note that a Graph can be any kind of Graph, including GraphWU, MultigraphBUD, and MultiplexBUT, among others.

---

**Code 11: NNDataPoint\_Graph\_REG element props.** The props section of generator code in `_NNDataPoint_Graph_REG.gen.m` defines the properties to be used in `NNDataPoint_Graph_REG`.

---

```

1 %% iprops!
2
3 %% iprop! ①
4 G (data, item) is a graph.
5 %%%% isettings!
6 'Graph'
7
8 %% iprop!
9 SUB (data, item) is a subject.
10 %%%% isettings!
11 'Subject'
12
13 %% iprop!
14 TARGET_IDS (parameter, stringlist) is a list of variable-of-interest IDs to
    be used as the class targets.

```

① defines the Graph element which contains its corresponding adjacency matrix.



Code 12: **NNDataPoint\_Graph\_REG element tests**. The tests section from the element generator `_NNDataPoint_Graph_REG.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 %% iexcluded_props!
4 [NNDataPoint_Graph_REG.G NNDataPoint_Graph_REG.SUB]
5
6 %% itest!
7 %%% iname! ①
8 Construct the data point with the adjacency matrix derived from its weighted
   undirected graph (GraphWU)
9 %%% icode!
10 % ensure the example data is generated
11 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
   REG CON XLS' filesep 'atlas.xlsx'])
12     test_NNDataPoint_CON_REG % create example files
13 end
14
15 % Load BrainAtlas
16
17 ...
18
19 % Analysis CON WU ②
20 a_WU = AnalyzeEnsemble_CON_WU( ...
21     'GR', gr ...
22 );
23
24 a_WU.memorize('G_DICT'); ③
25
26 % create item lists of NNDataPoint_Graph_REG
27 it_list = cellfun(@(g, sub) ④ NNDataPoint_Graph_REG( ...
28     'ID', sub.get('ID'), ...
29     'G', g, ...
30     'SUB', sub, ...
31     'TARGET_IDS', sub.get('VOI_DICT').get('KEYS')), ...
32     ⑤ a_WU.get('G_DICT').get('IT_LIST'), gr.get('SUB_DICT').get('IT_LIST')
   , ...
33     'UniformOutput', false);
34
35 % create NNDataPoint_Graph_REG DICT items
36 dp_list = IndexedDictionary(...
37     'IT_CLASS', 'NNDataPoint_Graph_REG', ...
38     'IT_LIST', it_list ...
39 );
40
41 % create a NNDataset containing the NNDataPoint_Graph_REG DICT
42 d = NNDataset( ...
43     'DP_CLASS', 'NNDataPoint_Graph_REG', ...
44     'DP_DICT', dp_list ...
45 );
46
47 % Check whether the content of input for a single data point matches ⑥
48 for index = 1:1:gr.get('SUB_DICT').get('LENGTH')
49     individual_input = d.get('DP_DICT').get('IT', index).get('INPUT');
50     known_input = a_WU.get('G_DICT').get('IT', index).get('A');

```

① tests with the GraphWU element which contains weighted undirected adjacency matrix.

② and ③ create the `AnalyzeEnsemble_CON_WU` element and then memorize its graph dictionary `G_DICT`.

④ and ⑤ creates the `NNDataPoint_Graph_REG` element and use the Graph from `G_DICT`.

⑥ tests whether the value of each input from the data point matches the graph's adjacency matrix.

```

51
52     assert(isequal(individual_input, known_input), ...
53         [BRAPH2.STR 'NNDataPoint_Graph_REG:' BRAPH2.FAIL_TEST], ...
54         'NNDataPoint_Graph_REG does not construct the dataset correctly. The
           input value is not derived correctly.' ...
55     )
56 end
57
58 %%% itest!
59 %%% iname! ⑦
60 Construct the data point with the adjacency matrix derived from its binary
    undirected multigraph with fixed densities (MultigraphBUD)
61 %%% icode!
62 % ensure the example data is generated
63 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
    REG CON XLS' filesep 'atlas.xlsx'])
64     test_NNDataPoint_CON_REG % create example files
65 end
66
67 % Load BrainAtlas
68
69 ...
70
71 % Analysis CON WU
72 densities = 0:25:100;
73
74 a_BUD = ⑧AnalyzeEnsemble_CON_BUD( ...
75     'DENSITIES', densities, ...
76     'GR', gr ...
77 );
78
79 a_BUD.memorize('G_DICT');
80
81 % create item lists of NNDataPoint_Graph_REG
82 it_list = cellfun(@(g, sub) ⑨ NNDataPoint_Graph_REG( ...
83     'ID', sub.get('ID'), ...
84     'G', g, ...
85     'SUB', sub, ...
86     'TARGET_IDS', sub.get('VOI_DICT').get('KEYS')), ...
87     ⑩a_BUD.get('G_DICT').get('IT_LIST'), gr.get('SUB_DICT').get('IT_LIST'
88     ),...
89     'UniformOutput', false);
89
90 % create NNDataPoint_Graph_REG DICT items
91 dp_list = IndexedDictionary(...
92     'IT_CLASS', 'NNDataPoint_Graph_REG', ...
93     'IT_LIST', it_list ...
94 );
95
96 ...
97
98 %%% itest!
99 %%% iname! ⑪
100 Construct the data point with the adjacency matrix derived from its
    multiplex weighted undirected graph (MultiplexWU)
101 %%% icode!
102 % ensure the example data is generated
103 if ~isfile([fileparts(which('SubjectCON_FUN_MP')) filesep 'Example data
    CON_FUN_MP XLS' filesep 'atlas.xlsx'])

```

⑦ tests with the MultigraphBUD element which contains the adjacency matrix of binary undirected graph at fixed densities.

⑧—⑩ create the NNDataPoint\_Graph\_REG element and use the Graph from AnalyzeEnsemble\_CON\_BUD.

⑪ tests with the MultiplexWU element which contains the adjacency matrix of weighted undirected multiplex.

```

104     test_SubjectCON_FUN_MP % create example files
105 end
106
107 % Load BrainAtlas
108
109 ...
110
111 % Analysis CON FUN MP WU
112 a_WU = (12)AnalyzeEnsemble_CON_FUN_MP_WU( ...
113     'GR', gr ...
114 );
115
116 a_WU.memorize('G_DICT');
117
118 % create item lists of NNDataPoint_Graph_REG
119 it_list = cellfun(@(g, sub) (13)NNDataPoint_Graph_REG( ...
120     'ID', sub.get('ID'), ...
121     'G', g, ...
122     'SUB', sub, ...
123     'TARGET_IDS', sub.get('VOI_DICT').get('KEYS')), ...
124     (14)a_WU.get('G_DICT').get('IT_LIST'), gr.get('SUB_DICT').get('IT_LIST')
125     , ...
126     'UniformOutput', false);
127
128 % create NNDataPoint_Graph_REG DICT items
129 dp_list = IndexedDictionary(...
130     'IT_CLASS', 'NNDataPoint_Graph_REG', ...
131     'IT_LIST', it_list ...
132 );
133 ...
134
135 %% itest!
136 %% iname! (15)
137 Example script for binary undirected graph (MultigraphBUT) using
138     connectivity data
139
140 %% icode!
141 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
142     REG CON XLS' filesep 'atlas.xlsx'])
143     test_NNDataPoint_CON_REG % create example files
144 end
145 example_NNCV_CON_BUT_REG
146
147 %% itest!
148 %% iname! (16)
149 Example script for binary undirected multiplex at fixed densities (
150     MultiplexBUD) using connectivity data and functional data
151
152 %% icode!
153 if ~isfile([fileparts(which('NNDataPoint_CON_FUN_MP_REG')) filesep 'Example
154     data NN REG CON_FUN_MP XLS' filesep 'atlas.xlsx'])
155     test_NNDataPoint_CON_FUN_MP_REG % create example files
156 end
157 example_NNCV_CON_FUN_MP_BUD_REG
158
159 %% itest!
160 %% iname! (17)
161 Example script for binary undirected multiplex at fixed thresholds (
162     MultiplexBUT) using connectivity data and functional data

```

(12)–(14) creates the  
NNDataPoint\_Graph\_REG  
element and use the Graph from  
AnalyzeEnsemble\_CON\_BUD.

(15) tests with the MultigraphBUT  
element with the simulated connectivity  
data.

(16) tests with the MultiplexBUD  
element with the simulated connectivity  
and functional data.

(17) tests with the MultiplexBUT  
element with the simulated connectivity  
and functional data.

```
156 %%%% icode!
157 if ~isfile([fileparts(which('NNDataPoint_CON_FUN_MP_REG')) filesep 'Example
    data NN REG CON_FUN_MP XLS' filesep 'atlas.xlsx'])
158     test_NNDataPoint_CON_FUN_MP_REG % create example files
159 end
160 example_NNCV_CON_FUN_MP_BUT_REG
```

---

### *Graph data point for classification (NNDataPoint\_Graph\_CLA)*

Now we implement NNDataPoint\_Graph\_CLA based on previous codes NNDataPoint\_CON\_CLA. This neural network data point with graphs utilizes the adjacency matrix extracted from the derived graph of the subject. The modified parts of the code are highlighted.

**Code 13: NNDataPoint\_Graph\_CLA element header.** The header section of the generator code in \_NNDataPoint\_Graph\_CLA.gen.m provides the general information about the NNDataPoint\_Graph\_CLA element.

---

```

1 %% iheader!
2 NNDataPoint_Graph_CLA < NNDataPoint (dp, graph classification data point) is
  a data point for classification with a graph.
3
4 %%% idescription!
5 A data point for classification with a graph (NNDataPoint_Graph_CLA)
6 contains both input and target for neural network analysis.
7 The input is the value of the adjacency matrix extracted from the derived
  graph of the subject.
8 The target is obtained from the variables of interest of the subject.

```

---

**Code 14: NNDataPoint\_Graph\_CLA element props update.** The props\_update section of the generator code in \_NNDataPoint\_Graph\_CLA.gen.m updates the properties of the NNDataPoint\_Graph\_CLA element. This defines the core properties of the data point.

---

```

1 %% iprops_update!
2
3 %%% iprop!
4 NAME (constant, string) is the name of a data point for classification with
  a graph.
5 %%% idefault!
6 'NNDataPoint_Graph_CLA'
7
8 %%% iprop!
9 DESCRIPTION (constant, string) is the description of a data point for
  classification with a graph.
10 %%% idefault!
11 'A data point for classification with a graph (NNDataPoint_Graph_CLA)
  contains both input and target for neural network analysis. The input
  is the value of the adjacency matrix extracted from the derived graph
  of the subject. The target is obtained from the variables of interest
  of the subject.'
12
13 %%% iprop!
14 TEMPLATE (parameter, item) is the template of a data point for
  classification with a graph.
15 %%% isettings!
16 'NNDataPoint_Graph_CLA'
17
18 %%% iprop!
19 ID (data, string) is a few-letter code for a data point for classification
  with a graph.
20 %%% idefault!

```

---

```

21 'NNDataPoint_Graph_CLA ID'
22
23 %% iprop!
24 LABEL (metadata, string) is an extended label of a data point for
    classification with a graph.
25 %%%% idefault!
26 'NNDataPoint_Graph_CLA label'
27
28 %% iprop!
29 NOTES (metadata, string) are some specific notes about a data point for
    classification with a graph.
30 %%%% idefault!
31 'NNDataPoint_Graph_CLA notes'
32
33 %% iprop!
34 INPUT (result, cell) is the input value for this data point.
35 %%%% icalculate!
36 value = dp.get('G').get('A'); ①
37
38 %% iprop!
39 TARGET (result, cell) is the target value for this data point.
40 %%%% icalculate!
41 value = dp.get('TARGET_IDS');

```

---

① extracts the adjacency matrix from a Graph element as the input for this data point. Note that a Graph can be any kind of Graph, including GraphWU, MultigraphBUD, and MultiplexBUT, among others.

**Code 15: NNDataPoint\_Graph\_CLA element props.** The props section of generator code in `_NNDataPoint_Graph_CLA.gen.m` defines the properties to be used in `NNDataPoint_Graph_CLA`.

---

```

1 %% iprops!
2
3 %% iprop! ①
4 G (data, item) is a graph.
5 %%%% issettings!
6 'Graph'
7
8 %% iprop!
9 TARGET_IDS (parameter, stringlist) is a list of variable-of-interest IDs to
    be used as the class targets.

```

---

① defines the Graph element which contains its corresponding adjacency matrix.

Code 16: **NNDataPoint\_Graph\_CLA element tests**. The tests section from the element generator `_NNDataPoint_Graph_CLA.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 %% iexcluded_props!
4 [NNDataPoint_Graph_CLA.G]
5
6 %% itest!
7 %%% iname! ①
8 Construct the data point with the adjacency matrix derived from its weighted
   undirected graph (GraphWU)
9 %%% icode!
10 % ensure the example data is generated
11 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
   CLA CON XLS' filesep 'atlas.xlsx'])
12     test_NNDataPoint_CON_CLA % create example files
13 end
14
15 % Load BrainAtlas
16
17 ...
18
19 % Analysis CON WU
20 a_WU1 = ②AnalyzeEnsemble_CON_WU( ...
21     'GR', gr1 ...
22 );
23
24 a_WU2 = AnalyzeEnsemble_CON_WU( ...
25     'TEMPLATE', a_WU1, ...
26     'GR', gr2 ...
27 );
28
29 a_WU1.memorize('G_DICT');
30 a_WU2.memorize('G_DICT');
31
32 % create item lists of NNDataPoint_Graph_CLA
33 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
34 it_list1 = cellfun(@(x) ③NNDataPoint_Graph_CLA( ...
35     'ID', x.get('ID'), ...
36     'G', x, ...
37     'TARGET_IDS', {group_folder_name}), ...
38     ④a_WU1.get('G_DICT').get('IT_LIST'), ...
39     'UniformOutput', false);
40
41 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
42 it_list2 = cellfun(@(x) NNDataPoint_Graph_CLA( ...
43     'ID', x.get('ID'), ...
44     'G', x, ...
45     'TARGET_IDS', {group_folder_name}), ...
46     a_WU2.get('G_DICT').get('IT_LIST'), ...
47     'UniformOutput', false);
48
49 % create NNDataPoint_Graph_CLA DICT items
50
51 ...

```

① tests with the GraphWU element which contains weighted undirected adjacency matrix.

②—④ create Analyzensemble\_CON\_WU and use its G\_DICT to initialize NNDataPoint\_Graph\_CLA.

```

52
53 %% itest!
54 %%% iname! ⑤
55 Construct the data point with the adjacency matrix derived from its binary
    undirected multigraph with fixed densities (MultigraphBUD)
56 %%% icode!
57 % ensure the example data is generated
58 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
    CLA CON XLS' filesep 'atlas.xlsx'])
59     test_NNDataPoint_CON_CLA % create example files
60 end
61
62 % Load BrainAtlas
63
64 ...
65
66 % Analysis CON WU
67 densities = 0:25:100;
68
69 a_BUD1 = ⑥AnalyzeEnsemble_CON_BUD( ...
70     'DENSITIES', densities, ...
71     'GR', gr1 ...
72 );
73
74 a_BUD2 = AnalyzeEnsemble_CON_BUD( ...
75     'TEMPLATE', a_BUD1, ...
76     'GR', gr2 ...
77 );
78
79 a_BUD1.memorize('G_DICT');
80 a_BUD2.memorize('G_DICT');
81
82 % create item lists of NNDataPoint_Graph_CLA
83 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
84 it_list1 = cellfun(@(x) ⑦NNDataPoint_Graph_CLA( ...
85     'ID', x.get('ID'), ...
86     'G', x, ...
87     'TARGET_IDS', {group_folder_name}), ...
88     ⑧a_BUD1.get('G_DICT').get('IT_LIST'), ...
89     'UniformOutput', false);
90
91 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
92 it_list2 = cellfun(@(x) NNDataPoint_Graph_CLA( ...
93     'ID', x.get('ID'), ...
94     'G', x, ...
95     'TARGET_IDS', {group_folder_name}), ...
96     a_BUD2.get('G_DICT').get('IT_LIST'), ...
97     'UniformOutput', false);
98
99 % create NNDataPoint_Graph_CLA DICT items
100 dp_list1 = IndexedDictionary(...
101     'IT_CLASS', 'NNDataPoint_Graph_CLA', ...
102     'IT_LIST', it_list1 ...
103 );
104
105 dp_list2 = IndexedDictionary(...
106     'IT_CLASS', 'NNDataPoint_Graph_CLA', ...
107     'IT_LIST', it_list2 ...
108 );
109

```

⑤ tests with the MultigraphBUD element which contains binary undirected adjacency matrix at fixed densities.

⑥—⑧ create Analyzensemble\_CON\_BUD and use its G\_DICT to initialize NNDataPoint\_Graph\_CLA.



```

110 % create a NNdataset containing the NNDataPoint_Graph_CLA DICT
111
112 ...
113
114 %% itest!
115 %%% iname! ⑨
116 Construct the data point with the adjacency matrix derived from its
      multiplex weighted undirected graph (MultiplexWU)
117 %%% icode!
118 % ensure the example data is generated
119 if ~isfile([fileparts(which('SubjectCON_FUN_MP')) filesep 'Example data
      CON_FUN_MP XLS' filesep 'atlas.xlsx'])
120     test_SubjectCON_FUN_MP % create example files
121 end
122
123 % Load BrainAtlas
124
125 ...
126
127 % Analysis CON FUN MP WU
128 a_WU1 = ⑩ AnalyzeEnsemble_CON_FUN_MP_WU( ...
129     'GR', gr1 ...
130 );
131
132 a_WU2 = AnalyzeEnsemble_CON_FUN_MP_WU( ...
133     'TEMPLATE', a_WU1, ...
134     'GR', gr2 ...
135 );
136
137 a_WU1.memorize('G_DICT');
138 a_WU2.memorize('G_DICT');
139
140 % create item lists of NNDataPoint_Graph_CLA
141 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
142 it_list1 = cellfun(@(x) ⑪ NNDataPoint_Graph_CLA( ...
143     'ID', x.get('ID'), ...
144     'G', x, ...
145     'TARGET_IDS', {group_folder_name}), ...
146     ⑫ a_WU1.get('G_DICT').get('IT_LIST'), ...
147     'UniformOutput', false);
148
149 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
150 it_list2 = cellfun(@(x) NNDataPoint_Graph_CLA( ...
151     'ID', x.get('ID'), ...
152     'G', x, ...
153     'TARGET_IDS', {group_folder_name}), ...
154     a_WU2.get('G_DICT').get('IT_LIST'), ...
155     'UniformOutput', false);
156
157 % create NNDataPoint_Graph_CLA DICT items
158
159 ...
160
161 %% itest!
162 %%% iname! ⑬
163 Example script for weighted undirected graph (GraphWU) using connectivity
      data
164 %%% icode!
165 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN

```

⑨ tests with the MultiplexWU element which contains weighted undirected adjacency matrix from multiplex graph.

⑩—⑫ create Analyzensemble\_CON\_FUN\_MP\_WU and use its G\_DICT to initialize NNDataPoint\_Graph\_CLA.

⑬ tests with the GraphWU element with simulated data.

```

        CLA CON XLS' filesep 'atlas.xlsx'])
166     test_NNDataPoint_CON_CLA % create example files
167 end
168 example_NNCV_CON_WU_CLA
169
170 %%% itest!
171 %%% iname! 14
172 Example script for binary undirected graph at fixed densities (MultigraphBUD
    ) using connectivity data
173 %%% icode!
174 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
    CLA CON XLS' filesep 'atlas.xlsx'])
175     test_NNDataPoint_CON_CLA % create example files
176 end
177 example_NNCV_CON_BUD_CLA
178
179 %%% itest!
180 %%% iname! 15
181 Example script for weighted undirected multiplex (MultiplexWU) using
    connectivity data and functional data
182 %%% icode!
183 if ~isfile([fileparts(which('NNDataPoint_CON_FUN_MP_CLA')) filesep 'Example
    data NN CLA CON_FUN_MP XLS' filesep 'atlas.xlsx'])
184     test_NNDataPoint_CON_FUN_MP_CLA % create example files
185 end
186 example_NNCV_CON_FUN_MP_WU_CLA

```

---

14 tests with the MultigraphBUD element with simulated data.

15 tests with the MultiplexWU element with simulated data.

## Implementation of a data point with graph measures

### Graph measure data point for regression (NNDataPoint\_Measure\_REG)

Now we implement NNDataPoint\_Measure\_REG based on previous codes NNDataPoint\_Graph\_REG. This neural network data point utilizes graph measures obtained from the adjacency matrix from the derived graph of the subject. The modified parts of the code are highlighted.

#### Code 17: NNDataPoint\_Measure\_REG element

**header.** The header section of the generator code in `_NNDataPoint_Measure_REG.gen.m` provides the general information about the NNDataPoint\_Measure\_REG element.

---

```

1 %% iheader!
2 NNDataPoint_Measure_REG < NNDataPoint (dp, measure regression data point) is
   a data point for regression with graph measures.
3
4 %%% idescription!
5 A data point for regression with graph measures (NNDataPoint_Measure_REG)
6 contains both input and target for neural network analysis.
7 The input is the value of the graph measures (e.g. Degree, DegreeAv, and
   Distance),
8 calculated from the derived graph of the subject.
9 The target is obtained from the variables of interest of the subject.

```

---

#### Code 18: NNDataPoint\_Measure\_REG element props up-

**date.** The props\_update section of the generator code in `_NNDataPoint_Measure_REG.gen.m` updates the properties of the NNDataPoint\_Measure\_REG element. This defines the core properties of the data point.

---

```

1 %% iprops_update!
2
3 %%% iprop!
4 NAME (constant, string) is the name of a data point for regression with
   graph measures.
5 %%%% idefault!
6 'NNDataPoint_Measure_REG'
7
8 %%% iprop!
9 DESCRIPTION (constant, string) is the description of a data point for
   regression with graph measures.
10 %%%% idefault!
11 'A data point for regression with graph measures (NNDataPoint_Measure_REG)
   contains both input and target for neural network analysis. The input
   is the value of the graph measures (e.g. Degree, DegreeAv, and Distance
   ), calculated from the derived graph of the subject. The target is
   obtained from the variables of interest of the subject.'
12
13 %%% iprop!
14 TEMPLATE (parameter, item) is the template of a data point for regression
   with graph measures.
15 %%%% isettings!
16 'NNDataPoint_Measure_REG'

```

---

```

17
18 %% iprop!
19 ID (data, string) is a few-letter code for a data point for regression with
    graph measures.
20 %%%% idefault!
21 'NNDataPoint_Measure_REG ID'
22
23 %% iprop!
24 LABEL (metadata, string) is an extended label of a data point for regression
    with graph measures.
25 %%%% idefault!
26 'NNDataPoint_Measure_REG label'
27
28 %% iprop!
29 NOTES (metadata, string) are some specific notes about a data point for
    regression with graph measures.
30 %%%% idefault!
31 'NNDataPoint_Measure_REG notes'
32
33 %% iprop!
34 INPUT (result, cell) is the input value for this data point.
35 %%%% icalculate!
36 value = cellfun(@(m_class) dp.get('G').get('MEASURE', m_class).get('M'), dp.
    get('M_LIST'), 'UniformOutput', false); ①
37
38 %% iprop!
39 TARGET (result, cell) is the target value for this data point.
40 %%%% icalculate!
41 value = cellfun(@(x) dp.get('SUB').get('VOI_DICT').get('IT', x).get('V'), dp
    .get('TARGET_IDS'), 'UniformOutput', false);

```

① calculates the graph measures, specified with M\_LIST, from a Graph element for this data point. Note that a Graph can be any kind of Graph, including GraphWU, MultigraphBUD, and MultiplexBUT, among others.

---

**Code 19: NNDataPoint\_Measure\_REG element props.** The props section of generator code in `_NNDataPoint_Measure_REG.gen.m` defines the properties to be used in `NNDataPoint_Measure_REG`.

---

```

1 %% iprops!
2
3 %%%% iprop!
4 G (data, item) is a graph.
5 %%%% isettings!
6 'Graph'
7
8 %%%% iprop! ①
9 M_LIST (parameter, classlist) is a list of graph measure to be used as the
    input
10
11 %% iprop!
12 SUB (data, item) is a subject.
13 %%%% isettings!
14 'Subject'
15
16 %% iprop!
17 TARGET_IDS (parameter, stringlist) is a list of variable-of-interest IDs to
    be used as the class targets.

```

① defines the graph measure list which will be obtained as INPUT for this data point.

Code 20: **NNDataPoint\_Measure\_REG element tests**. The tests section from the element generator `_NNDataPoint_Measure_REG.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 %% iexcluded_props!
4 [NNDataPoint_Measure_REG.G NNDataPoint_Measure_REG.SUB]
5
6 %%% itest!
7 %%% iname!
8 Construct the data point with the adjacency matrix derived from its weighted
   undirected graph (GraphWU)
9 %%% icode!
10 % ensure the example data is generated
11 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
   REG CON XLS' filesep 'atlas.xlsx'])
12     test_NNDataPoint_CON_REG % create example files
13 end
14
15 % Load BrainAtlas
16
17 ...
18
19 % Analysis CON WU
20 a_WU = (1)AnalyzeEnsemble_CON_WU( ...
21     'GR', gr ...
22 );
23
24 a_WU.get('MEASUREENSEMBLE', 'Degree').get('M'); (2)
25 a_WU.get('MEASUREENSEMBLE', 'DegreeAv').get('M'); (3)
26 a_WU.get('MEASUREENSEMBLE', 'Distance').get('M'); (4)
27
28 % create item lists of NNDataPoint_Measure_REG
29 it_list = cellfun(@(g, sub) (5)NNDataPoint_Measure_REG( ...
30     'ID', sub.get('ID'), ...
31     'G', g, ...
32     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
33     'SUB', sub, ...
34     'TARGET_IDS', sub.get('VOI_DICT').get('KEYS')), ...
35     (6)a_WU.get('G_DICT').get('IT_LIST'), gr.get('SUB_DICT').get('IT_LIST')
36     , ...
37     'UniformOutput', false);
38
39 % create NNDataPoint_Measure_REG DICT items
40 dp_list = IndexedDictionary(...
41     'IT_CLASS', 'NNDataPoint_Measure_REG', ...
42     'IT_LIST', it_list ...
43 );
44 ...
45
46 %%% itest!
47 %%% iname! (7)
48 Construct the data point with the adjacency matrix derived from its binary
   undirected multigraph with fixed densities (MultigraphBUD)

```

(1)–(4) create a `AnalyzeEnsemble_CON_WU` and add various kinds of graph measure with the `GraphWU` element which contains weighted undirected adjacency matrix.

(5) and (6) use `AnalyzeEnsemble_CON_WU`'s `G_DICT` to set up a `NNDataPoint_Measure_REG`.

(7)–(10) tests various kinds of graph measure with the `MultigraphBUD`.

```

49 %%%% icode!
50 % ensure the example data is generated
51 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
    REG CON XLS' filesep 'atlas.xlsx'])
52     test_NNDataPoint_CON_REG % create example files
53 end
54
55 % Load BrainAtlas
56
57 ...
58
59 % Analysis CON WU
60 densities = 0:25:100;
61
62 a_BUD = (8)AnalyzeEnsemble_CON_BUD( ...
63     'DENSITIES', densities, ...
64     'GR', gr ...
65 );
66
67 a_BUD.get('MEASUREENSEMBLE', 'Degree').get('M');
68 a_BUD.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
69 a_BUD.get('MEASUREENSEMBLE', 'Distance').get('M');
70
71 % create item lists of NNDataPoint_Measure_REG
72 it_list = cellfun(@(g, sub) (9)NNDataPoint_Measure_REG( ...
73     'ID', sub.get('ID'), ...
74     'G', g, ...
75     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
76     'SUB', sub, ...
77     'TARGET_IDS', sub.get('VOI_DICT').get('KEYS')), ...
78     (10)a_BUD.get('G_DICT').get('IT_LIST'), gr.get('SUB_DICT').get('IT_LIST'
79     ),...
80     'UniformOutput', false);
81
82 % create NNDataPoint_CON_CLA_DICT items
83 dp_list = IndexedDictionary(...
84     'IT_CLASS', 'NNDataPoint_Measure_REG', ...
85     'IT_LIST', it_list ...
86 );
87 ...
88
89 %%% itest!
90 %%% iname! (11)
91 Construct the data point with the adjacency matrix derived from its
    multiplex weighted undirected graph (MultiplexWU)
92 %%%% icode!
93 % ensure the example data is generated
94 if ~isfile([fileparts(which('SubjectCON_FUN_MP')) filesep 'Example data
    CON_FUN_MP XLS' filesep 'atlas.xlsx'])
95     test_SubjectCON_FUN_MP % create example files
96 end
97
98 % Load BrainAtlas
99
100 ...
101
102 % Analysis CON FUN MP WU
103 a_WU = (12)AnalyzeEnsemble_CON_FUN_MP_WU( ...

```

(11) — (14) tests various kinds of graph measure with the MultiplexWU.

```

104     'GR', gr ...
105 );
106
107 % To be added the multiplex measures
108 a_WU.get('MEASUREENSEMBLE', 'Degree').get('M');
109 a_WU.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
110 a_WU.get('MEASUREENSEMBLE', 'Distance').get('M');
111
112 % create item lists of NNDataPoint_Measure_REG
113 it_list = cellfun(@(g, sub) (13) NNDataPoint_Measure_REG( ...
114     'ID', sub.get('ID'), ...
115     'G', g, ...
116     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
117     'SUB', sub, ...
118     'TARGET_IDS', sub.get('VOI_DICT').get('KEYS')), ...
119     (14) a_WU.get('G_DICT').get('IT_LIST'), gr.get('SUB_DICT').get('IT_LIST')
120     , ...
121     'UniformOutput', false);
122 % create NNDataPoint_Measure_REG DICT items
123
124 ...
125
126 %%% itest!
127 %%% iname! (15)
128 Example script for weighted undirected graph (GraphWU) using connectivity
    data
129 %%% icode!
130 if ~isfile([fileparts(which('NNDataPoint_CON_REG')) filesep 'Example data NN
    REG CON XLS' filesep 'atlas.xlsx'])
131     test_NNDataPoint_CON_REG % create example files
132 end
133 example_NNCV_CON_WU_M_REG
134
135 %%% itest!
136 %%% iname! (16)
137 Example script for weighted undirected multiplex (MultiplexWU) using
    connectivity data and functional data
138 %%% icode!
139 if ~isfile([fileparts(which('NNDataPoint_CON_FUN_MP_REG')) filesep 'Example
    data NN REG CON_FUN_MP XLS' filesep 'atlas.xlsx'])
140     test_NNDataPoint_CON_FUN_MP_REG % create example files
141 end
142 example_NNCV_CON_FUN_MP_WU_M_REG

```

(15) tests various kinds of graph measure with the GraphWU using example data.

(16) tests various kinds of graph measure with the MultiplexWU using example data.

### *Graph measure data point for classification (NNDataPoint\_Measure\_CLA)*

Now we implement NNDataPoint\_Measure\_CLA based on previous codes NNDataPoint\_Graph\_CLA. This neural network data point utilizes graph measures obtained from the adjacency matrix from the derived graph of the subject. The modified parts of the code are highlighted.

#### Code 21: NNDataPoint\_Measure\_CLA element

**header.** The header section of the generator code in `_NNDataPoint_Measure_CLA.gen.m` provides the general information about the NNDataPoint\_Measure\_CLA element.

---

```

1 %% iheader!
2 NNDataPoint_Measure_CLA < NNDataPoint (dp, measure classification data point
   ) is a data point for classification with graph measures.
3
4 %%% idescription!
5 A data point for classification with graph measures (NNDataPoint_Measure_CLA
   )
6 contains both input and target for neural network analysis.
7 The input is the value of the graph measures (e.g. Degree, DegreeAv, and
   Distance),
8 calculated from the derived graph of the subject.
9 The target is obtained from the variables of interest of the subject.

```

---

#### Code 22: NNDataPoint\_Measure\_CLA element props up-

**date.** The props\_update section of the generator code in `_NNDataPoint_Measure_CLA.gen.m` updates the properties of the NNDataPoint\_Measure\_CLA element. This defines the core properties of the data point.

---

```

1 %% iprops_update!
2
3 %%% iprop!
4 NAME (constant, string) is the name of a data point for classification with
   graph measures.
5 %%% idefault!
6 'NNDataPoint_Measure_CLA'
7
8 %%% iprop!
9 DESCRIPTION (constant, string) is the description of a data point for
   classification with graph measures.
10 %%% idefault!
11 'A data point for classification with graph measures (
   NNDataPoint_Measure_CLA) contains both input and target for neural
   network analysis. The input is the value of the graph measures (e.g.
   Degree, DegreeAv, and Distance), calculated from the derived graph of
   the subject. The target is obtained from the variables of interest of
   the subject.'
12
13 %%% iprop!
14 TEMPLATE (parameter, item) is the template of a data point for
   classification with graph measures.
15 %%% isettings!
16 'NNDataPoint_Measure_CLA'
17

```

---



```

18 %%% iprop!
19 ID (data, string) is a few-letter code for a data point for classification
    with graph measures.
20 %%% idefault!
21 'NNDataPoint_Measure_CLA ID'
22
23 %%% iprop!
24 LABEL (metadata, string) is an extended label of a data point for
    classification with graph measures.
25 %%% idefault!
26 'NNDataPoint_Measure_CLA label'
27
28 %%% iprop!
29 NOTES (metadata, string) are some specific notes about a data point for
    classification with graph measures.
30 %%% idefault!
31 'NNDataPoint_Measure_CLA notes'
32
33 %%% iprop!
34 INPUT (result, cell) is the input value for this data point.
35 %%% icalculate!
36 value = cellfun(@(m_class) dp.get('G').get('MEASURE', m_class).get('M'), dp.
    get('M_LIST'), 'UniformOutput', false); ①
37
38 %%% iprop!
39 TARGET (result, cell) is the target value for this data point.
40 %%% icalculate!
41 value = dp.get('TARGET_IDS');

```

---

**Code 23: NNDataPoint\_Measure\_CLA element props.** The props section of generator code in `_NNDataPoint_Measure_CLA.gen.m` defines the properties to be used in `NNDataPoint_Measure_CLA`.

---

```

1 %%% iprops!
2
3 %%% iprop!
4 G (data, item) is a graph.
5 %%% isettings!
6 'Graph'
7
8 %%% iprop! ①
9 M_LIST (parameter, classlist) is a list of graph measure to be used as the
    input
10
11 %%% iprop!
12 TARGET_IDS (parameter, stringlist) is a list of variable-of-interest IDs to
    be used as the class targets.

```

---

① calculates or extract the graph measures, which are specified with `M_LIST` from a Graph element for this data point. Note that a Graph can be any kind of Graph, including `GraphWU`, `MultigraphBUD`, and `MultiplexBUT`, among others.

① defines the graph measure list which will be obtained as `INPUT` for this data point.

Code 24: **NNDataPoint\_Measure\_CLA element tests.** The tests section from the element generator `_NNDataPoint_Measure_CLA.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 %%% iexcluded_props!
4 [NNDataPoint_Measure_CLA.G]
5
6 %%% itest!
7 %%% iname!
8 Construct the data point with the graph measures derived from its weighted
   undirected graph (GraphWU)
9 %%% icode!
10 % ensure the example data is generated
11 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
   CLA CON XLS' filesep 'atlas.xlsx'])
12     test_NNDataPoint_CON_CLA % create example files
13 end
14
15 % Load BrainAtlas
16
17 ...
18
19 % Analysis CON WU
20 a_WU1 = (1)AnalyzeEnsemble_CON_WU( ...
21     'GR', gr1 ...
22 );
23
24 a_WU2 = AnalyzeEnsemble_CON_WU( ...
25     'TEMPLATE', a_WU1, ...
26     'GR', gr2 ...
27 );
28
29 a_WU1.get('MEASUREENSEMBLE', 'Degree').get('M'); (2)
30 a_WU1.get('MEASUREENSEMBLE', 'DegreeAv').get('M'); (3)
31 a_WU1.get('MEASUREENSEMBLE', 'Distance').get('M'); (4)
32
33 a_WU2.get('MEASUREENSEMBLE', 'Degree').get('M');
34 a_WU2.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
35 a_WU2.get('MEASUREENSEMBLE', 'Distance').get('M');
36
37 % create item lists of NNDataPoint_Measure_CLA
38 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
39 it_list1 = cellfun(@(x) (5)NNDataPoint_Measure_CLA( ...
40     'ID', x.get('ID'), ...
41     'G', x, ...
42     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
43     'TARGET_IDS', {group_folder_name}), ...
44     (6)a_WU1.get('G_DICT').get('IT_LIST'), ...
45     'UniformOutput', false);
46
47 % create NNDataPoint_Measure_CLA DICT items
48
49 ...
50
51 %%% itest!

```

(1)–(4) create a `AnalyzeEnsemble_CON_WU` and add various kinds of graph measure with the `GraphWU` element which contains weighted undirected adjacency matrix. (5) and (6) use `AnalyzeEnsemble_CON_WU`'s `G_DICT` to set up a `NNDataPoint_Measure_CLA`.

```

52 %%%% iiname! ⑦
53 Construct the data point with the graph measures derived from its binary
    undirected multigraph with fixed densities (MultigraphBUD)
54 %%%% icode!
55 % ensure the example data is generated
56 if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
    CLA CON XLS' filesep 'atlas.xlsx'])
57     test_NNDataPoint_CON_CLA % create example files
58 end
59
60 % Load BrainAtlas
61
62 ...
63
64 % Analysis CON WU
65 densities = 0:25:100;
66
67 a_BUD1 = ⑧ AnalyzeEnsemble_CON_BUD( ...
68     'DENSITIES', densities, ...
69     'GR', gr1 ...
70 );
71
72 a_BUD2 = AnalyzeEnsemble_CON_BUD( ...
73     'TEMPLATE', a_BUD1, ...
74     'GR', gr2 ...
75 );
76
77 a_BUD1.get('MEASUREENSEMBLE', 'Degree').get('M'); ⑨
78 a_BUD1.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
79 a_BUD1.get('MEASUREENSEMBLE', 'Distance').get('M');
80
81 a_BUD2.get('MEASUREENSEMBLE', 'Degree').get('M');
82 a_BUD2.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
83 a_BUD2.get('MEASUREENSEMBLE', 'Distance').get('M');
84
85 % create item lists of NNDataPoint_Measure_CLA
86 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
87 it_list1 = cellfun(@(x) ⑩ NNDataPoint_Measure_CLA( ...
88     'ID', x.get('ID'), ...
89     'G', x, ...
90     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
91     'TARGET_IDS', {group_folder_name}), ...
92     a_BUD1.get('G_DICT').get('IT_LIST'), ...
93     'UniformOutput', false);
94
95 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
96 it_list2 = cellfun(@(x) NNDataPoint_Measure_CLA( ...
97     'ID', x.get('ID'), ...
98     'G', x, ...
99     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
100     'TARGET_IDS', {group_folder_name}), ...
101     a_BUD2.get('G_DICT').get('IT_LIST'), ...
102     'UniformOutput', false);
103
104 % create NNDataPoint_Measure_CLA items
105 dp_list1 = IndexedDictionary(...
106     'IT_CLASS', 'NNDataPoint_Measure_CLA', ...
107     'IT_LIST', it_list1 ...
108 );
109

```

⑦—⑩ tests various kinds of graph measure with the MultigraphBUD.

```

110 ...
111
112 %%% itest!
113 %%% iname! ⑪
114 Construct the data point with the graph measures derived from its multiplex
    weighted undirected graph (MultiplexWU)
115 %%% icode!
116 % ensure the example data is generated
117 if ~isfile([fileparts(which('SubjectCON_FUN_MP')) filesep 'Example data
    CON_FUN_MP XLS' filesep 'atlas.xlsx'])
118     test_SubjectCON_FUN_MP % create example files
119 end
120
121 % Load BrainAtlas
122
123 ...
124
125 % Analysis CON FUN MP WU
126 a_WU1 = ⑫ AnalyzeEnsemble_CON_FUN_MP_WU( ...
127     'GR', gr1 ...
128 );
129
130 a_WU2 = AnalyzeEnsemble_CON_FUN_MP_WU( ...
131     'TEMPLATE', a_WU1, ...
132     'GR', gr2 ...
133 );
134
135 % To be added the multiplex measures
136 a_WU1.get('MEASUREENSEMBLE', 'Degree').get('M'); ⑬
137 a_WU1.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
138 a_WU1.get('MEASUREENSEMBLE', 'Distance').get('M');
139
140 a_WU2.get('MEASUREENSEMBLE', 'Degree').get('M');
141 a_WU2.get('MEASUREENSEMBLE', 'DegreeAv').get('M');
142 a_WU2.get('MEASUREENSEMBLE', 'Distance').get('M');
143
144 % create item lists of NNDataPoint_Graph_CLA
145 [~, group_folder_name] = fileparts(im_gr1.get('DIRECTORY'));
146 it_list1 = cellfun(@(x) ⑭ NNDataPoint_Measure_CLA( ...
147     'ID', x.get('ID'), ...
148     'G', x, ...
149     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
150     'TARGET_IDS', {group_folder_name}), ...
151     a_WU1.get('G_DICT').get('IT_LIST'), ...
152     'UniformOutput', false);
153
154 [~, group_folder_name] = fileparts(im_gr2.get('DIRECTORY'));
155 it_list2 = cellfun(@(x) NNDataPoint_Measure_CLA( ...
156     'ID', x.get('ID'), ...
157     'G', x, ...
158     'M_LIST', {'Degree' 'DegreeAv' 'Distance'}, ...
159     'TARGET_IDS', {group_folder_name}), ...
160     a_WU2.get('G_DICT').get('IT_LIST'), ...
161     'UniformOutput', false);
162
163 % create NNDataPoint_Measure_CLA_DICT items
164 dp_list1 = IndexedDictionary(...
165     'IT_CLASS', 'NNDataPoint_Measure_CLA', ...
166     'IT_LIST', it_list1 ...

```

⑪—⑭ tests various kinds of graph measure with the MultigraphBUD.

```

167     );
168
169     ...
170
171     %% itest!
172     %%% iname! (15)
173     Example script for binary undirected graph at fixed densities (GraphBUD)
        using connectivity data
174     %%% icode!
175     if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
        CLA CON XLS' filesep 'atlas.xlsx'])
176         test_NNDataPoint_CON_CLA % create example files
177     end
178     example_NNCV_CON_BUD_M_CLA
179
180     %% itest!
181     %%% iname! (16)
182     Example script for binary undirected graph at fixed thresholds (
        MultigraphBUT) using connectivity data
183     %%% icode!
184     if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
        CLA CON XLS' filesep 'atlas.xlsx'])
185         test_NNDataPoint_CON_CLA % create example files
186     end
187     example_NNCV_CON_BUT_M_CLA
188
189     %% itest!
190     %%% iname! (17)
191     Example script for binary undirected graph at fixed densities (MultigraphBUD
        ) using connectivity data
192     %%% icode!
193     if ~isfile([fileparts(which('NNDataPoint_CON_CLA')) filesep 'Example data NN
        CLA CON XLS' filesep 'atlas.xlsx'])
194         test_NNDataPoint_CON_CLA % create example files
195     end
196     example_NNCV_CON_BUD_M_CLA
197
198     %% itest!
199     %%% iname! (18)
200     Example script for binary undirected multiplex at fixed densities (
        MultiplexBUD) using connectivity data and functional data
201     %%% icode!
202     if ~isfile([fileparts(which('NNDataPoint_CON_FUN_MP_CLA')) filesep 'Example
        data NN CLA CON_FUN_MP XLS' filesep 'atlas.xlsx'])
203         test_NNDataPoint_CON_FUN_MP_CLA % create example files
204     end
205     example_NNCV_CON_FUN_MP_BUD_M_CLA
206
207     %% itest!
208     %%% iname! (19)
209     Example script for binary undirected multiplex at fixed thresholds (
        MultiplexBUT) using connectivity data and functional data
210     %%% icode!
211     if ~isfile([fileparts(which('NNDataPoint_CON_FUN_MP_CLA')) filesep 'Example
        data NN CLA CON_FUN_MP XLS' filesep 'atlas.xlsx'])
212         test_NNDataPoint_CON_FUN_MP_CLA % create example files
213     end
214     example_NNCV_CON_FUN_MP_BUT_M_CLA

```

(15) tests various kinds of graph measure with the MultigraphBUD using example connectivity data.

(16) tests various kinds of graph measure with the MultigraphBUT using example data.

(17) tests various kinds of graph measure with the MultigraphBUD using example connectivity data.

(18) tests various kinds of graph measure with the MultiplexBUD using example functional data.

(19) tests various kinds of graph measure with the MultiplexBUT using example functional data.