

ANEXO V Detección Dinámica de Comunidades

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José Pedro Manzano Patrón

Lenguaje: Matlab

Este anexo muestra el código necesario para, a partir de las series temporales de cada región cerebral (ROI_timeseries):

- Generar el inventariado temporal, de tamaño w y solapamiento temporal *overlapping*.
- Generar las matrices de conectividad asociada a cada ventana temporal, con la métrica solicitada (Pearson, correlación parcial o *distance correlation*).
- Generar 3 modelos nulos diferentes: un temporal que aleatoriza las series temporales, un modelo nulo nodal que aleatoriza los enlaces entre nodos y un modelo temporal que aleatoriza el orden de las ventanas temporales de las matrices de conectividad.
- Detección de comunidades para cada ventana temporal.
- Aplicación de métodos de consenso sobre las comunidades.
- Principales métricas de la comunidad y del grafo para cada ventana temporal.

El código hace llamadas a subrutinas de forma continua. Por favor, asegúrese que ha descargado e incluido en la ruta todo el código que aparece en el repositorio de Github.

Si tuviese algún problema, no dude en contactar a josepman@ucm.es

```
In [ ]: clear all
        close all

        % Selecciona la ruta correcta de su ordenador
        ROI_timeseries = csvread('/Users/hose/Desktop/TFM_TECI/Code/ts_con_CSF.csv',1,1);

        nNodes = size(ROI_timeseries,2);
        w = 20; % tamaño ventana
        overlapping = 0.8;
        l = size(ROI_timeseries,1); % número de nodos

        % A partir de las series temporales de cada nodo, genera las ventanas
        % temporales de tamaño w y solapamiento temporal = overlapping
        sliding_windows = gen_sliding_window(ROI_timeseries, w, overlapping);

        % Create the connectivity matrix
        metric_rule = 'corr'; % 'corr', 'dist_corr', 'partial_corr', 'cross_corr', 'MI'
        connectivity_matrix = gen_connectivity_matrix(sliding_windows, metric_rule);
```

```

% NULL-MODELS
for i=1:size(ROI_timeseries,2);
    tseries_null_model(:,i) = tseries_rand(ROI_timeseries(:,i));
end
tseries_null_model_sw = gen_sliding_window(tseries_null_model, w, overlapping);
tseries_null_model_conn = gen_connectivity_matrix(tseries_null_model_sw, metric_rule);

for i=1:size(connectivity_matrix,3)
    nodal_null_model(:,i) = randmio_und_connected(connectivity_matrix(:,i), 50);
end

temporal_null_model_conn = temporal_rand(connectivity_matrix); % Random

% Para GN, descomente las siguientes líneas:
% gamma = 2;
% omega = 1;
% conn_mat = cell(1,size(connectivity_matrix,3));
% for i=1:size(connectivity_matrix,3)
%     conn_mat{1,i} = connectivity_matrix(:,i);
% end
% [B,twom] = multicat(conn_mat,gamma,omega); % M
%
models = {connectivity_matrix,tseries_null_model_conn,temporal_null_model_conn };

for j=1:length(models);
    connectivity_matrix = models{3};
    % Get the communities
    n = 100;
    final_partition = zeros(nNodes,size(connectivity_matrix,3));
    final_Q = zeros(1,size(connectivity_matrix,3));
    thr_nodal = cell(1,size(connectivity_matrix,3));
    quality = zeros(1,size(connectivity_matrix,3));
    consensus = zeros(nNodes,size(connectivity_matrix,3));
    consensus_simm = zeros(1,size(connectivity_matrix,3));
    agreement_nmi = zeros(1,size(connectivity_matrix,3));
    agreement_z = zeros(1,size(connectivity_matrix,3));
    SWP = zeros(1,size(connectivity_matrix,3));
    delta_c = zeros(1,size(connectivity_matrix,3));
    delta_L = zeros(1,size(connectivity_matrix,3));
    persistence_LMC = cell(1,size(connectivity_matrix,3));

    for i=1:size(connectivity_matrix,3)
        % final partition is created from consensus clustering. See
        % community_detection script
        [final_partition(:,i), Q, thr_nodal{1,i}, quality(i), consensus(:,i), consensus_simm(:,i), agreement_nmi(i), agreement_z(i), SWP(i), delta_c(i), delta_L(i), persistence_LMC{i})] = GN(connectivity_matrix(:,i), n);

        %Similarity between both methods
    end
end

```

```

        agreement_nmi(i) = normalized_mutual_information(consensus(:,i), final_partition(:,i));
        agreement_z(i) = zrand(consensus(:,i), final_partition(:,i));

        [SWP(i), delta_C(i), delta_L(i)] = small_world_propensity(connectivity_matrix(:,i), delta_C(i), delta_L(i));

        persistence_LMC{1,i} = sig_lmc(final_partition(:,1), connectivity_matrix(:,1));
    end

    %%%%%
    % Improvement of multilayer partitions of categorical networks
    final_partition = postprocess_categorical_multilayer(final_partition, size(final_partition, 2));

    % reorders nodes and layers to emphasize persistent structure in an unordered multilayer network
    [final_partition_sorted, s1, s2] = sort_categorical(final_partition);
    pers = categorical_persistence(final_partition_sorted);
    %%%%%

    % agreement between windows:
    agreement_w_nmi = zeros(size(connectivity_matrix, 3), size(connectivity_matrix, 3));
    agreement_w_z = agreement_w_nmi;
    for i=1:size(connectivity_matrix, 3)
        for j=1:size(connectivity_matrix, 3)
            agreement_w_nmi(i, j) = normalized_mutual_information(final_partition(:, i), final_partition(:, j));
            agreement_w_z(i, j) = zrand(final_partition(:, i), final_partition(:, j));
        end
    end

    % Coeff of changes of comm. per node
    flexib = flexibility(final_partition, 'cat');

    % Cohesion (if a node change, other also change):
    [Cij, node_cohesion, node_disjoint, node_flexibility, strength_cohesion, commChanges, commStrength] = cohesion(final_partition);
    % Promiscuity (fraction of all communities in which a node participates at least once)
    P = promiscuity(final_partition);

    % Persistence
    persist = persistence(final_partition);

    % El resto de métricas se pueden extraer con la función bct_measures:
    thr = 0.5; % para option='binary'
    option = 'weighted';

    G{i} = bct_measures(connectivity_matrix, option, thr);
end

```