

# Predicting coefficients for graphs

## Contents

<b>1</b>	<b>Motivation</b>	<b>3</b>
<b>2</b>	<b>Features</b>	<b>3</b>
2.1	Feature Categories . . . . .	3
<b>3</b>	<b>Intra-loop modelling</b>	<b>4</b>
3.1	Feature reduction . . . . .	4
<b>4</b>	<b>Predicting <math>l</math> from <math>p &lt; l</math> loop orders</b>	<b>4</b>
4.1	Approach 1: Timeseries approach . . . . .	4
4.1.1	Feature reduction . . . . .	4
4.1.2	Shap Analysis . . . . .	4
4.2	Approach 2: General $l$ learning . . . . .	4
4.2.1	Feature reduction . . . . .	4
4.2.2	Shap Analysis . . . . .	4
4.3	Ensemble learning . . . . .	4
<b>A</b>	<b>Complete Feature Descriptions</b>	<b>4</b>

## **Abstract**

In this set of notes we do the following.

- **denominator graphs:**
  - **Intra-loop modelling** - ability to model within loop level
  - **Modelling higher-loop** based on previous loop order
  - **Shap value analysis** - what is driving the decision
- **f-graphs**

# 1 Motivation

## 2 Features

This section provides a comprehensive overview of all graph features used in our analysis. The features are extracted using two main tools: `fgraph_features_cli2.py` for comprehensive graph analysis and `motif_features_cli.py` for motif-specific features.

### 2.1 Feature Categories

The features are organized into the following categories:

- **Basic:** Fundamental graph properties (nodes, edges, degrees, density, clustering)
- **Connectivity:** Path-based metrics (diameter, radius, shortest paths, components)
- **Centrality:** Node importance measures (betweenness, closeness, eigenvector)
- **Core:** K-core decomposition metrics
- **Robustness:** Vulnerability measures (articulation points, bridges)
- **Cycles:** Cycle counting features
- **Spectral Laplacian:** Laplacian matrix eigenvalues and related metrics
- **NetLSD:** Network Laplacian Spectral Descriptor features
- **Planarity:** Planar embedding properties
- **Symmetry:** Graph automorphism features
- **Community:** Community detection metrics
- **Motifs\_3\_4:** 3-node and 4-node motif counts
- **Motifs\_5:** 5-node motif counts
- **Motifs\_4:** 4-node induced subgraph counts

- **Spectral Adjacency:** Adjacency matrix spectrum features
- **TDA:** Topological Data Analysis features (persistent homology)
- **Normalized variants:** Size-normalized versions of many features

A complete list of all 179 features with their descriptions and interpretations can be found in Appendix A.

### 3 Intra-loop modelling

#### 3.1 Feature reduction

### 4 Predicting $l$ from $p < l$ loop orders

#### 4.1 Approach 1: Timeseries approach

##### 4.1.1 Feature reduction

##### 4.1.2 Shap Analysis

#### 4.2 Approach 2: General $l$ learning

##### 4.2.1 Feature reduction

##### 4.2.2 Shap Analysis

#### 4.3 Ensemble learning

## A Complete Feature Descriptions

This appendix provides a comprehensive list of all  $\mu$  graph features with their descriptions and interpretations. Due to the large number of features, the table is split across multiple pages for readability.

Feature Name	Group	Description	Interpretation
Basic_num_nodes	Basic	Total number of nodes in the graph	The larger this number the bigger the graph is
Basic_num_edges	Basic	Total number of edges in the graph	The larger this number the more connected the graph is
Basic_min_degree	Basic	Minimum degree among all nodes	The larger this number the more connected the least connected node is
Basic_max_degree	Basic	Maximum degree among all nodes	The larger this number the more connected the most connected node is
Basic_avg_degree	Basic	Average degree across all nodes	The larger this number the more connected the graph is on average
Basic_degree_std	Basic	Standard deviation of node degrees	The larger this number the more unequal the node connections are
Basic_degree_skew	Basic	Skewness of degree distribution	Positive values mean more high-degree nodes; negative means more low-degree nodes
Basic_density	Basic	Graph density (edges/max_possible_edges)	The larger this number the more densely connected the graph is
Basic_edge_to_node_ratio	Basic	Ratio of edges to nodes	The larger this number the more edges per node the graph has
Basic_degree_entropy	Basic	Shannon entropy of degree distribution	The larger this number the more diverse the node degrees are
Assortativity_degree	Basic	Degree assortativity coefficient	Positive values mean similar-degree nodes connect; negative means opposite-degree nodes connect
Clustering_mean	Basic	Average local clustering coefficient	The larger this number the more clustered/triangular the graph is
Clustering_q10	Basic	10th percentile of clustering coefficients	The larger this number the more clustered the least clustered nodes are
Clustering_q50	Basic	50th percentile (median) of clustering coefficients	The larger this number the more clustered the typical node is
Clustering_q90	Basic	90th percentile of clustering coefficients	The larger this number the more clustered the most clustered nodes are
Clustering_frac_zero	Basic	Fraction of nodes with zero clustering	The larger this number the more tree-like the graph is
Clustering_frac_one	Basic	Fraction of nodes with clustering = 1	The larger this number the more clique-like the graph is
Degree_gini	Basic	Gini coefficient of degree distribution	The larger this number the more unequal the node degrees are

Feature Name	Group	Description	Interpretation
Connectivity_is_connected	Connectivity	Whether graph is connected (True/False)	True means all nodes can reach each other; False means graph is fragmented
Connectivity_num_components	Connectivity	Number of connected components	The larger this number the more fragmented the graph is
Connectivity_diameter	Connectivity	Graph diameter (longest shortest path)	The larger this number the more spread out the graph is
Connectivity_radius	Connectivity	Graph radius (minimum eccentricity)	The larger this number the more spread out the graph is
Connectivity_avg_shortest_path_length	Connectivity	Average shortest path length	The larger this number the more spread out the graph is
Connectivity_wiener_index	Connectivity	Sum of all shortest path lengths	The larger this number the more spread out the graph is
Eff_diameter_p90	Connectivity	90th percentile effective diameter	The larger this number the more spread out the graph is
Ecc_mean	Connectivity	Mean eccentricity of nodes	The larger this number the more spread out the graph is
Ecc_q90	Connectivity	90th percentile eccentricity	The larger this number the more spread out the graph is
Centrality_betweenness_mean	Centrality	Mean betweenness centrality	The larger this number the more nodes act as bridges/connectors
Centrality_betweenness_max	Centrality	Maximum betweenness centrality	The larger this number the more important the most central node is
Centrality_betweenness_std	Centrality	Standard deviation of betweenness centrality	The larger this number the more unequal the node importance is
Centrality_betweenness_skew	Centrality	Skewness of betweenness centrality distribution	Positive values mean few very important nodes; negative means many moderately important nodes
Centrality_closeness_mean	Centrality	Mean closeness centrality	The larger this number the more centrally located nodes are on average
Centrality_closeness_max	Centrality	Maximum closeness centrality	The larger this number the more centrally located the most central node is
Centrality_closeness_std	Centrality	Standard deviation of closeness centrality	The larger this number the more unequal the node centrality is
Centrality_closeness_skew	Centrality	Skewness of closeness centrality distribution	Positive values mean few very central nodes; negative means many moderately central nodes
Centrality_eigenvector_mean	Centrality	Mean eigenvector centrality	The larger this number the more nodes are connected to important nodes
Centrality_eigenvector_max	Centrality	Maximum eigenvector centrality	The larger this number the more important the most influential node is
Centrality_eigenvector_std	Centrality	Standard deviation of eigenvector centrality	The larger this number the more unequal the node influence is

Feature Name	Group	Description	Interpretation
Centrality_eigenvector_skew	Centrality	<i>Skewness of eigenvector centrality distribution</i>	Positive values mean few very influential nodes; negative means many moderately influential nodes
Core_max_core_index	Core	<i>Maximum k-core index</i>	The larger this number the more tightly connected the densest core is
Core_core_index_mean	Core	<i>Mean k-core index</i>	The larger this number the more tightly connected nodes are on average
Robust_articulation_points	Robustness	<i>Number of articulation points (cut vertices)</i>	The larger this number the more vulnerable the graph is to fragmentation
Robust_bridge_count	Robustness	<i>Number of bridges (cut edges)</i>	The larger this number the more vulnerable the graph is to fragmentation
Cycle_num_cycles_len_5	Cycles	<i>Number of cycles of length 5</i>	The larger this number the more 5-cycles the graph contains
Cycle_num_cycles_len_6	Cycles	<i>Number of cycles of length 6</i>	The larger this number the more 6-cycles the graph contains
Spectral_algebraic_connectivity	Spectral_Laplacian	<i>Second smallest Laplacian eigenvalue (Fiedler value)</i>	The larger this number the more connected the graph is
Spectral_spectral_gap	Spectral_Laplacian	<i>Difference between first two Laplacian eigenvalues</i>	The larger this number the more well-connected the graph is
Spectral_laplacian_mean	Spectral_Laplacian	<i>Mean of Laplacian eigenvalues</i>	The larger this number the more connected the graph is on average
Spectral_laplacian_std	Spectral_Laplacian	<i>Standard deviation of Laplacian eigenvalues</i>	The larger this number the more varied the connectivity patterns are
Spectral_laplacian_skew	Spectral_Laplacian	<i>Skewness of Laplacian eigenvalue distribution</i>	Positive values mean few highly connected components; negative means many moderately connected components
Spectral_lap_eig_0	Spectral_Laplacian	<i>Smallest Laplacian eigenvalue</i>	Always 0 for connected graphs; larger values indicate more disconnected components
Spectral_lap_eig_1	Spectral_Laplacian	<i>Second smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
Spectral_lap_eig_2	Spectral_Laplacian	<i>Third smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
Spectral_lap_eig_3	Spectral_Laplacian	<i>Fourth smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
Spectral_lap_eig_4	Spectral_Laplacian	<i>Fifth smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
Spectral_lap_eig_5	Spectral_Laplacian	<i>Sixth smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
Spectral_lap_eig_6	Spectral_Laplacian	<i>Seventh smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
Spectral_lap_eig_7	Spectral_Laplacian	<i>Eighth smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is

Feature Name	Group	Description	Interpretation
<b>Spectral_lap_eig_8</b>	Spectral.Laplacian	<i>Ninth smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
<b>Spectral_lap_eig_9</b>	Spectral.Laplacian	<i>Tenth smallest Laplacian eigenvalue</i>	The larger this number the more connected the graph is
<b>Kirchhoff_index</b>	Spectral.Laplacian	<i>Kirchhoff index (sum of resistance distances)</i>	The larger this number the more spread out the graph is
<b>NetLSD_mean</b>	NetLSD	<i>Mean NetLSD signature</i>	The larger this number the more complex the graph structure is
<b>NetLSD_std</b>	NetLSD	<i>Standard deviation of NetLSD signature</i>	The larger this number the more varied the graph structure is
<b>NetLSD_q10</b>	NetLSD	<i>10th percentile of NetLSD signature</i>	The larger this number the more complex the simplest parts are
<b>NetLSD_q90</b>	NetLSD	<i>90th percentile of NetLSD signature</i>	The larger this number the more complex the most complex parts are
<b>Planarity_num_faces</b>	Planarity	<i>Number of faces in planar embedding</i>	The larger this number the more complex the planar structure is
<b>Planarity_face_size_mean</b>	Planarity	<i>Mean face size in planar embedding</i>	The larger this number the larger the typical face is
<b>Planarity_face_size_max</b>	Planarity	<i>Maximum face size in planar embedding</i>	The larger this number the larger the biggest face is
<b>Symmetry_automorphism_group_order</b>	Symmetry	<i>Order of automorphism group</i>	The larger this number the more symmetric the graph is
<b>Symmetry_num_orbits</b>	Symmetry	<i>Number of node orbits under automorphisms</i>	The larger this number the more diverse the node roles are
<b>Symmetry_orbit_size_max</b>	Symmetry	<i>Maximum orbit size</i>	The larger this number the more nodes share the same role
<b>Comm_modularity</b>	Community	<i>Modularity of best community partition</i>	The larger this number the more clearly separated the communities are
<b>Comm_count</b>	Community	<i>Number of communities found</i>	The larger this number the more fragmented the graph is
<b>Comm_size_max</b>	Community	<i>Size of largest community</i>	The larger this number the more dominant the largest community is
<b>Comm_size_gini</b>	Community	<i>Gini coefficient of community sizes</i>	The larger this number the more unequal the community sizes are
<b>Comm_internal_edge_frac</b>	Community	<i>Fraction of edges within communities</i>	The larger this number the more internally connected communities are
<b>Motif_triangles</b>	Motifs.3.4	<i>Number of triangles (3-cliques)</i>	The larger this number the more triangular structures the graph has
<b>Motif_wedges</b>	Motifs.3.4	<i>Number of wedges (2-paths)</i>	The larger this number the more path-like structures the graph has



Feature Name	Group	Description	Interpretation
Motif.4_cycles	Motifs.3.4	Number of 4-cycles	The larger this number the more square-like structures the graph has
Motif.4_cliques	Motifs.3.4	Number of 4-cliques ( $K_4$ )	The larger this number the more tightly connected 4-node groups the graph has
Motif.triangle_edge_incidence_mean	Motifs.3.4	Mean triangles per edge	The larger this number the more triangles each edge participates in
Motif.triangle_edge_incidence_std	Motifs.3.4	Standard deviation of triangles per edge	The larger this number the more varied edge participation in triangles is
Motif.square_clustering_proxy	Motifs.3.4	Tendency to form 4-cycles relative to 2-paths	The larger this number the more square-like the graph structure is
Motif.triangle_edge_incidence_median	Motifs.3.4	Median triangles per edge	The larger this number the more triangles the typical edge participates in
Motif.triangle_edge_incidence_90	Motifs.3.4	90th percentile triangles per edge	The larger this number the more triangles the most triangular edges participate in
Motif.triangle_edge_frac_zero	Motifs.3.4	Fraction of edges with zero triangles	The larger this number the more tree-like the graph is
Motif.triangle_edge_frac_ge2	Motifs.3.4	Fraction of edges with $\geq 2$ triangles	The larger this number the more clustered the graph is
Motif.induced_K1_3	Motifs.3.4	Number of induced $K_{1,3}$ (star) subgraphs	The larger this number the more star-like structures the graph has
Motif.induced_P4	Motifs.3.4	Number of induced $P_4$ (path) subgraphs	The larger this number the more path-like structures the graph has
Motif.induced_C4	Motifs.3.4	Number of induced $C_4$ (cycle) subgraphs	The larger this number the more cycle-like structures the graph has
Motif.induced_TailedTriangle	Motifs.3.4	Number of induced tailed triangle subgraphs	The larger this number the more tailed triangle structures the graph has
Motif.induced_Diamond	Motifs.3.4	Number of induced diamond subgraphs	The larger this number the more diamond structures the graph has
Motif.induced_K4	Motifs.3.4	Number of induced $K_4$ (clique) subgraphs	The larger this number the more tightly connected 4-node groups the graph has
Motif.induced_connected_per_4	Motifs.3.4	Fraction of 4-node subsets that are connected	The larger this number the more connected 4-node groups are
Motif.5_cycles	Motifs.5	Number of 5-cycles	The larger this number the more 5-sided cycle structures the graph has
Motif.5_cliques	Motifs.5	Number of 5-cliques ( $K_5$ )	The larger this number the more tightly connected 5-node groups the graph has
Motif.induced5_g_0_5	Motifs.5	Number of induced 5-node graphlet $g_{.0}$	The larger this number the more $g_{.0}$ structures the graph has
Motif.induced5_g_1_5	Motifs.5	Number of induced 5-node graphlet $g_{.1}$	The larger this number the more $g_{.1}$ structures the graph has

Feature Name	Group	Description	Interpretation
Motif.induced5_g_2_5	Motifs_5	Number of induced 5-node graphlet g_2	The larger this number the more g_2 structures the graph has
Motif.induced5_g_3_5	Motifs_5	Number of induced 5-node graphlet g_3	The larger this number the more g_3 structures the graph has
Motif.induced5_g_4_5	Motifs_5	Number of induced 5-node graphlet g_4	The larger this number the more g_4 structures the graph has
Motif.induced5_g_5_5	Motifs_5	Number of induced 5-node graphlet g_5	The larger this number the more g_5 structures the graph has
Motif.induced5_g_6_5	Motifs_5	Number of induced 5-node graphlet g_6	The larger this number the more g_6 structures the graph has
Motif.induced5_g_7_5	Motifs_5	Number of induced 5-node graphlet g_7	The larger this number the more g_7 structures the graph has
Motif.induced5_g_8_5	Motifs_5	Number of induced 5-node graphlet g_8	The larger this number the more g_8 structures the graph has
Motif.induced5_g_9_5	Motifs_5	Number of induced 5-node graphlet g_9	The larger this number the more g_9 structures the graph has
Motif.induced5_g_10_5	Motifs_5	Number of induced 5-node graphlet g_10	The larger this number the more g_10 structures the graph has
Motif.induced5_g_11_5	Motifs_5	Number of induced 5-node graphlet g_11	The larger this number the more g_11 structures the graph has
Motif.induced5_g_12_5	Motifs_5	Number of induced 5-node graphlet g_12	The larger this number the more g_12 structures the graph has
Motif.induced5_g_13_5	Motifs_5	Number of induced 5-node graphlet g_13	The larger this number the more g_13 structures the graph has
Motif.induced5_g_14_5	Motifs_5	Number of induced 5-node graphlet g_14	The larger this number the more g_14 structures the graph has
Motif.induced5_g_15_5	Motifs_5	Number of induced 5-node graphlet g_15	The larger this number the more g_15 structures the graph has
Motif.induced5_g_16_5	Motifs_5	Number of induced 5-node graphlet g_16	The larger this number the more g_16 structures the graph has
Motif.induced5_g_17_5	Motifs_5	Number of induced 5-node graphlet g_17	The larger this number the more g_17 structures the graph has
Motif.induced5_g_18_5	Motifs_5	Number of induced 5-node graphlet g_18	The larger this number the more g_18 structures the graph has
Motif.induced5_g_20_5	Motifs_5	Number of induced 5-node graphlet g_20	The larger this number the more g_20 structures the graph has
Motif.induced.connected.per.5	Motifs_5	Fraction of 5-node subsets that are connected	The larger this number the more connected 5-node groups are
Adjacency.energy	Spectral.Adjacency	Sum of absolute eigenvalues of adjacency matrix	The larger this number the more energetic/vibrant the graph is

Feature Name	Group	Description	Interpretation
Adjacency_estrada_index	Spectral-Adjacency	Sum of exponentials of eigenvalues	The larger this number the more communicable the graph is
Adjacency_moment_2	Spectral-Adjacency	Second moment of adjacency eigenvalues	The larger this number the more spread out the adjacency spectrum is
Adjacency_moment_3	Spectral-Adjacency	Third moment of adjacency eigenvalues	Positive values mean more high-frequency components; negative means more low-frequency components
Adjacency_moment_4	Spectral-Adjacency	Fourth moment of adjacency eigenvalues	The larger this number the more peaked the adjacency spectrum is
Spectral_laplacian_heat_trace_t0.1	Spectral-Laplacian	Laplacian heat trace at $t=0.1$	The larger this number the more heat spreads quickly through the graph
Spectral_laplacian_heat_trace_t1.0	Spectral-Laplacian	Laplacian heat trace at $t=1.0$	The larger this number the more heat spreads through the graph
Spectral_laplacian_heat_trace_t5.0	Spectral-Laplacian	Laplacian heat trace at $t=5.0$	The larger this number the more heat spreads through the graph
TDA_H0_count	TDA	Number of $H0$ homology features (connected components)	The larger this number the more disconnected components the graph has
TDA_H0_total_persistence	TDA	Total persistence of $H0$ features	The larger this number the more persistent the connectivity structure is
TDA_H0_mean_persistence	TDA	Mean persistence of $H0$ features	The larger this number the more stable the connectivity structure is
TDA_H0_max_persistence	TDA	Maximum persistence of $H0$ features	The larger this number the more stable the most persistent component is
TDA_H0_persistence_entropy	TDA	Entropy of $H0$ persistence distribution	The larger this number the more diverse the persistence values are
TDA_H0_mean_birth	TDA	Mean birth time of $H0$ features	The larger this number the later components typically appear
TDA_H0_mean_death	TDA	Mean death time of $H0$ features	The larger this number the later components typically disappear
TDA_H1_count	TDA	Number of $H1$ homology features (cycles)	The larger this number the more cyclic structures the graph has
TDA_H1_total_persistence	TDA	Total persistence of $H1$ features	The larger this number the more persistent the cyclic structure is
TDA_H1_mean_persistence	TDA	Mean persistence of $H1$ features	The larger this number the more stable the cyclic structure is
TDA_H1_max_persistence	TDA	Maximum persistence of $H1$ features	The larger this number the more stable the most persistent cycle is
TDA_H1_persistence_entropy	TDA	Entropy of $H1$ persistence distribution	The larger this number the more diverse the cycle persistence values are
TDA_H1_mean_birth	TDA	Mean birth time of $H1$ features	The larger this number the later cycles typically appear

Feature Name	Group	Description	Interpretation
TDA.H1_mean_death	TDA	Mean death time of $H1$ features	The larger this number the later cycles typically disappear
TDA.Betti0_at_q25	TDA	Betti number $\beta_0$ at 25th percentile filtration	The larger this number the more components exist at low filtration levels
TDA.Betti0_at_q50	TDA	Betti number $\beta_0$ at 50th percentile filtration	The larger this number the more components exist at medium filtration levels
TDA.Betti0_at_q75	TDA	Betti number $\beta_0$ at 75th percentile filtration	The larger this number the more components exist at high filtration levels
TDA.Betti1_at_q25	TDA	Betti number $\beta_1$ at 25th percentile filtration	The larger this number the more cycles exist at low filtration levels
TDA.Betti1_at_q50	TDA	Betti number $\beta_1$ at 50th percentile filtration	The larger this number the more cycles exist at medium filtration levels
TDA.Betti1_at_q75	TDA	Betti number $\beta_1$ at 75th percentile filtration	The larger this number the more cycles exist at high filtration levels
Basic_avg_degree_norm	Basic_Normalized	Average degree normalized by graph size	The larger this number the more connected the graph is relative to its size
Basic_degree_entropy_norm	Basic_Normalized	Degree entropy normalized by maximum possible	The larger this number the more diverse the node degrees are relative to maximum diversity
Connectivity_diameter_norm	Connectivity_Normalized	Diameter normalized by graph size	The larger this number the more spread out the graph is relative to its size
Connectivity_radius_norm	Connectivity_Normalized	Radius normalized by graph size	The larger this number the more spread out the graph is relative to its size
Wiener_mean_distance	Connectivity_Normalized	Mean distance normalized by Wiener index	The larger this number the more spread out the graph is relative to total distance
Motif_triangles_per_Cn3	Motifs_3_4_Normalized	Triangles normalized by $C(n,3)$	The larger this number the more triangular the graph is relative to maximum possible
Motif_4_cycles_per_Cn4	Motifs_3_4_Normalized	4-cycles normalized by $C(n,4)$	The larger this number the more square-like the graph is relative to maximum possible
Motif_4_cliques_per_Cn4	Motifs_3_4_Normalized	4-cliques normalized by $C(n,4)$	The larger this number the more tightly connected 4-node groups are relative to maximum possible
Motif_wedges_per_max	Motifs_3_4_Normalized	Wedges normalized by theoretical maximum	The larger this number the more path-like the graph is relative to maximum possible
Motif_induced_K1_3_per_Cn4	Motifs_3_4_Normalized	$K1,3$ normalized by $C(n,4)$	The larger this number the more star-like the graph is relative to maximum possible
Motif_induced_P4_per_Cn4	Motifs_3_4_Normalized	$P4$ normalized by $C(n,4)$	The larger this number the more path-like the graph is relative to maximum possible
Motif_induced_C4_per_Cn4	Motifs_3_4_Normalized	$C4$ normalized by $C(n,4)$	The larger this number the more cycle-like the graph is relative to maximum possible
Motif_induced_TailedTriangle_per_Cn4	Motifs_3_4_Normalized	Tailed triangle normalized by $C(n,4)$	The larger this number the more tailed triangle structures are relative to maximum possible

Feature Name	Group	Description	Interpretation
Motif.induced.Diamond_per_C4	Motifs.3.4.Normalized	Diamond normalized by $C(n,4)$	The larger this number the more diamond structures are relative to maximum possible
Motif.induced.K4_per_Cn4	Motifs.3.4.Normalized	$K4$ normalized by $C(n,4)$	The larger this number the more tightly connected 4-node groups are relative to maximum possible
Robust.articulation_points_per_node	Robustness.Normalized	Articulation points per node	The larger this number the more vulnerable the graph is to fragmentation per node
Robust.bridge_count_per_edge	Robustness.Normalized	Bridges per edge	The larger this number the more vulnerable the graph is to fragmentation per edge
Connectivity_num_components_per_node	Connectivity.Normalized	Components per node	The larger this number the more fragmented the graph is per node
Centrality_closeness_mean_norm	Centrality.Normalized	Mean closeness normalized by maximum	The larger this number the more centrally located nodes are on average relative to maximum
Centrality_closeness_max_norm	Centrality.Normalized	Max closeness normalized by maximum	The larger this number the more centrally located the most central node is relative to maximum
Spectral.algebraic_connectivity	Spectral.Normalized	Algebraic connectivity over average degree	The larger this number the more connected the graph is relative to its average connectivity
Spectral_spectral_gap_rel	Spectral.Normalized	Relative spectral gap	The larger this number the more well-connected the graph is relative to its connectivity
Spectral.laplacian_heat_trace.t0.1_per_node	Spectral.Normalized	Heat trace $t=0.1$ per node	The larger this number the more heat spreads quickly per node
Spectral.laplacian_heat_trace.t1.0_per_node	Spectral.Normalized	Heat trace $t=1.0$ per node	The larger this number the more heat spreads per node
Spectral.laplacian_heat_trace.t5.0_per_node	Spectral.Normalized	Heat trace $t=5.0$ per node	The larger this number the more heat spreads per node
Adjacency_energy_per_node	Spectral.Normalized	Adjacency energy per node	The larger this number the more energetic/vibrant the graph is per node
Adjacency_energy_over_fro	Spectral.Normalized	Adjacency energy over Frobenius norm	The larger this number the more energetic the graph is relative to its total energy
Adjacency_estrada_per_node	Spectral.Normalized	Estrada index per node	The larger this number the more communicable the graph is per node
log_Adjacency_estrada_per_node	Spectral.Normalized	Log Estrada index per node	The larger this number the more communicable the graph is per node (log scale)
Adjacency_moment_2_over_avg_deg	Spectral.Normalized	Second moment over average degree	The larger this number the more spread out the adjacency spectrum is relative to average connectivity
Adjacency_moment_3_over_avg_deg3	Spectral.Normalized	Third moment over average degree cubed	The larger this number the more high-frequency components are relative to connectivity cubed
Adjacency_moment_4_over_avg_deg4	Spectral.Normalized	Fourth moment over average degree to fourth	The larger this number the more peaked the adjacency spectrum is relative to connectivity to fourth power
Planarity_num_faces_over_upper_bound	Planarity.Normalized	Faces over theoretical upper bound	The larger this number the more complex the planar structure is relative to maximum possible

Feature Name	Group	Description	Interpretation
Planarity_face_size_mean_norm	Planarity_Normalized	Mean face size normalized	The larger this number the larger the typical face is relative to maximum possible
Symmetry_aut_size_log_over_log_n!	Symmetry_Normalized	Log automorphism size over log n!	The larger this number the more symmetric the graph is relative to maximum possible symmetry
Symmetry_num_orbits_per_node	Symmetry_Normalized	Orbits per node	The larger this number the more diverse the node roles are per node
Symmetry_orbit_size_max_per_node	Symmetry_Normalized	Max orbit size per node	The larger this number the more nodes share the same role per node
TDA_H0_count_per_node	TDA_Normalized	H0 features per node	The larger this number the more disconnected components exist per node
TDA_H0_total_persistence_over_diameter	TDA_Normalized	H0 persistence over diameter	The larger this number the more persistent the connectivity structure is relative to graph spread
TDA_H0_mean_persistence_over_diameter	TDA_Normalized	H0 mean persistence over diameter	The larger this number the more stable the connectivity structure is relative to graph spread
TDA_H0_max_persistence_over_diameter	TDA_Normalized	H0 max persistence over diameter	The larger this number the more stable the most persistent component is relative to graph spread
TDA_H0_mean_birth_over_diameter	TDA_Normalized	H0 mean birth over diameter	The larger this number the later components typically appear relative to graph spread
TDA_H0_mean_death_over_diameter	TDA_Normalized	H0 mean death over diameter	The larger this number the later components typically disappear relative to graph spread
TDA_H1_count_per_node	TDA_Normalized	H1 features per node	The larger this number the more cyclic structures exist per node
TDA_H1_total_persistence_over_diameter	TDA_Normalized	H1 persistence over diameter	The larger this number the more persistent the cyclic structure is relative to graph spread
TDA_H1_mean_persistence_over_diameter	TDA_Normalized	H1 mean persistence over diameter	The larger this number the more stable the cyclic structure is relative to graph spread
TDA_H1_max_persistence_over_diameter	TDA_Normalized	H1 max persistence over diameter	The larger this number the more stable the most persistent cycle is relative to graph spread
TDA_H1_mean_birth_over_diameter	TDA_Normalized	H1 mean birth over diameter	The larger this number the later cycles typically appear relative to graph spread
TDA_H1_mean_death_over_diameter	TDA_Normalized	H1 mean death over diameter	The larger this number the later cycles typically disappear relative to graph spread
TDA_Betti0_at_q25_per_node	TDA_Normalized	Betti0 at q25 per node	The larger this number the more components exist at low filtration levels per node
TDA_Betti1_at_q25_per_node	TDA_Normalized	Betti1 at q25 per node	The larger this number the more cycles exist at low filtration levels per node
TDA_Betti0_at_q50_per_node	TDA_Normalized	Betti0 at q50 per node	The larger this number the more components exist at medium filtration levels per node
TDA_Betti1_at_q50_per_node	TDA_Normalized	Betti1 at q50 per node	The larger this number the more cycles exist at medium filtration levels per node

Feature Name	Group	Description	Interpretation
TDA.Betti0.at.q75.per.node	TDA.Normalized	<i>Betti0 at q75 per node</i>	The larger this number the more components exist at high filtration levels per node
TDA.Betti1.at.q75.per.node	TDA.Normalized	<i>Betti1 at q75 per node</i>	The larger this number the more cycles exist at high filtration levels per node
Motif.induced.g.1.4	Motifs.4	<i>Number of induced Path4 (P4) subgraphs</i>	The larger this number the more path-like 4-node structures the graph has
Motif.induced.g.2.4	Motifs.4	<i>Number of induced Star4 (K1,3) subgraphs</i>	The larger this number the more star-like 4-node structures the graph has
Motif.induced.g.3.4	Motifs.4	<i>Number of induced Cycle4 (C4) subgraphs</i>	The larger this number the more cycle-like 4-node structures the graph has
Motif.induced.g.4.4	Motifs.4	<i>Number of induced TailedTriangle subgraphs</i>	The larger this number the more tailed triangle 4-node structures the graph has
Motif.induced.g.5.4	Motifs.4	<i>Number of induced Diamond subgraphs</i>	The larger this number the more diamond 4-node structures the graph has
Motif.induced.g.6.4	Motifs.4	<i>Number of induced Clique4 (K4) subgraphs</i>	The larger this number the more tightly connected 4-node groups the graph has
Motif.induced.g.1.4.per.Cn4	Motifs.4.Normalized	<i>Path4 normalized by C(n,4)</i>	The larger this number the more path-like 4-node structures are relative to maximum possible
Motif.induced.g.2.4.per.Cn4	Motifs.4.Normalized	<i>Star4 normalized by C(n,4)</i>	The larger this number the more star-like 4-node structures are relative to maximum possible
Motif.induced.g.3.4.per.Cn4	Motifs.4.Normalized	<i>Cycle4 normalized by C(n,4)</i>	The larger this number the more cycle-like 4-node structures are relative to maximum possible
Motif.induced.g.4.4.per.Cn4	Motifs.4.Normalized	<i>TailedTriangle normalized by C(n,4)</i>	The larger this number the more tailed triangle 4-node structures are relative to maximum possible
Motif.induced.g.5.4.per.Cn4	Motifs.4.Normalized	<i>Diamond normalized by C(n,4)</i>	The larger this number the more diamond 4-node structures are relative to maximum possible
Motif.induced.g.6.4.per.Cn4	Motifs.4.Normalized	<i>Clique4 normalized by C(n,4)</i>	The larger this number the more tightly connected 4-node groups are relative to maximum possible
Motif.induced5.g.0.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_0 normalized by C(n,5)</i>	The larger this number the more g_0 5-node structures are relative to maximum possible
Motif.induced5.g.1.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_1 normalized by C(n,5)</i>	The larger this number the more g_1 5-node structures are relative to maximum possible
Motif.induced5.g.2.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_2 normalized by C(n,5)</i>	The larger this number the more g_2 5-node structures are relative to maximum possible
Motif.induced5.g.3.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_3 normalized by C(n,5)</i>	The larger this number the more g_3 5-node structures are relative to maximum possible
Motif.induced5.g.4.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_4 normalized by C(n,5)</i>	The larger this number the more g_4 5-node structures are relative to maximum possible
Motif.induced5.g.5.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_5 normalized by C(n,5)</i>	The larger this number the more g_5 5-node structures are relative to maximum possible
Motif.induced5.g.6.5.per.Cn5	Motifs.5.Normalized	<i>5-node graphlet g_6 normalized by C(n,5)</i>	The larger this number the more g_6 5-node structures are relative to maximum possible

Feature Name	Group	Description	Interpretation
Motif.induced5_g_7_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_7 normalized by $C(n,5)$	The larger this number the more g_7 5-node structures are relative to maximum possible
Motif.induced5_g_8_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_8 normalized by $C(n,5)$	The larger this number the more g_8 5-node structures are relative to maximum possible
Motif.induced5_g_9_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_9 normalized by $C(n,5)$	The larger this number the more g_9 5-node structures are relative to maximum possible
Motif.induced5_g_10_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_10 normalized by $C(n,5)$	The larger this number the more g_10 5-node structures are relative to maximum possible
Motif.induced5_g_11_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_11 normalized by $C(n,5)$	The larger this number the more g_11 5-node structures are relative to maximum possible
Motif.induced5_g_12_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_12 normalized by $C(n,5)$	The larger this number the more g_12 5-node structures are relative to maximum possible
Motif.induced5_g_13_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_13 normalized by $C(n,5)$	The larger this number the more g_13 5-node structures are relative to maximum possible
Motif.induced5_g_14_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_14 normalized by $C(n,5)$	The larger this number the more g_14 5-node structures are relative to maximum possible
Motif.induced5_g_15_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_15 normalized by $C(n,5)$	The larger this number the more g_15 5-node structures are relative to maximum possible
Motif.induced5_g_16_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_16 normalized by $C(n,5)$	The larger this number the more g_16 5-node structures are relative to maximum possible
Motif.induced5_g_17_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_17 normalized by $C(n,5)$	The larger this number the more g_17 5-node structures are relative to maximum possible
Motif.induced5_g_18_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_18 normalized by $C(n,5)$	The larger this number the more g_18 5-node structures are relative to maximum possible
Motif.induced5_g_20_5_per_Cn5	Motifs_5_Normalized	5-node graphlet g_20 normalized by $C(n,5)$	The larger this number the more g_20 5-node structures are relative to maximum possible
Motif_5_cycles_per_Cn5	Motifs_5_Normalized	5-cycles normalized by $C(n,5)$	The larger this number the more 5-sided cycle structures are relative to maximum possible
Motif_5_cliques_per_Cn5	Motifs_5_Normalized	5-cliques normalized by $C(n,5)$	The larger this number the more tightly connected 5-node groups are relative to maximum possible
Motif_5_cycles_per_Kn	Motifs_5_Normalized	5-cycles normalized by complete graph	The larger this number the more 5-sided cycle structures are relative to complete graph
Spectral_kirchhoff_index	Spectral_Laplacian	Kirchhoff index (alternative name)	The larger this number the more spread out the graph is
Spectral_laplacian_heat_trace_t0.5	Spectral_Laplacian	Laplacian heat trace at $t=0.5$	The larger this number the more heat spreads through the graph
Spectral_laplacian_heat_trace_t2.0	Spectral_Laplacian	Laplacian heat trace at $t=2.0$	The larger this number the more heat spreads through the graph
Spectral_adjacency_energy	Spectral_Adjacency	Adjacency energy (alternative name)	The larger this number the more energetic/vibrant the graph is

  

Feature Name	Group	Description	Interpretation
Spectral_adjacency_estrada_index	Spectral_Adjacency	Adjacency Estrada index (alternative name)	The larger this number the more communicable the graph is
Spectral_adjacency_moment_2	Spectral_Adjacency	Adjacency second moment (alternative name)	The larger this number the more spread out the adjacency spectrum is
Spectral_adjacency_moment_3	Spectral_Adjacency	Adjacency third moment (alternative name)	Positive values mean more high-frequency components; negative means more low-frequency components
Spectral_adjacency_moment_4	Spectral_Adjacency	Adjacency fourth moment (alternative name)	The larger this number the more peaked the adjacency spectrum is