

Analysis

May 6, 2024

```
[1]: import scipy.io
import numpy as np
import networkx as nx
import os
import pandas as pd
from sklearn.cluster import SpectralClustering
# import community # Install using: pip install python-louvain
import matplotlib.pyplot as plt
import re
```

```
[2]: excel_file = 'fungal_networks_data/list_of_fungal_networks.xlsx'
fnd_df = pd.read_excel(excel_file, index_col='code')

filename = 'Pp_M_Tokyo_U_N_26h_1.txt'
substrate = fnd_df['substrate']
fnd_df
```

/opt/conda/lib/python3.10/site-packages/openpyxl/worksheet/_read_only.py:79:

UserWarning: Unknown extension is not supported and will be removed

for idx, row in parser.parse():

```
[2]: species size resources grazing interaction time point \
```

code						
Pp_M_Tokyo_U_N_26h_1	Pp	M	Tokyo	U	N	26h
Pp_M_Tokyo_U_N_26h_2	Pp	M	Tokyo	U	N	26h
Pp_M_Tokyo_U_N_26h_3	Pp	M	Tokyo	U	N	26h
Pp_M_Tokyo_U_N_26h_4	Pp	M	Tokyo	U	N	26h
Pp_M_Tokyo_U_N_26h_5	Pp	M	Tokyo	U	N	26h
...
Pv_M_I_U_Hf_18d_3	Pv	M	I	U	Hf	18d
Pv_M_I_U_Hf_22d_3	Pv	M	I	U	Hf	22d
Pv_M_I_U_Hf_26d_3	Pv	M	I	U	Hf	26d
Pv_M_I_U_Hf_38d_3	Pv	M	I	U	Hf	38d
Pv_M_I_U_Hf_48d_3	Pv	M	I	U	Hf	48d

```
repeat include substrate -K2 comments \
```

code					
Pp_M_Tokyo_U_N_26h_1	1	Y	A	-k2	NaN

Pp_M_Tokyo_U_N_26h_2	2	Y	A	-k2	NaN
Pp_M_Tokyo_U_N_26h_3	3	NaN	A	full	NaN
Pp_M_Tokyo_U_N_26h_4	4	Y	A	-k2	NaN
Pp_M_Tokyo_U_N_26h_5	5	NaN	A	full	NaN
...
Pv_M_I_U_Hf_18d_3	3	Y	S	-k2	NaN
Pv_M_I_U_Hf_22d_3	3	Y	S	-k2	NaN
Pv_M_I_U_Hf_26d_3	3	Y	S	-k2	NaN
Pv_M_I_U_Hf_38d_3	3	Y	S	-k2	NaN
Pv_M_I_U_Hf_48d_3	3	Y	S	-k2	NaN

citation

code

Pp_M_Tokyo_U_N_26h_1	Tero, A., Takagi, S., Saigusa, T., Ito, K., Be...
Pp_M_Tokyo_U_N_26h_2	Tero, A., Takagi, S., Saigusa, T., Ito, K., Be...
Pp_M_Tokyo_U_N_26h_3	Tero, A., Takagi, S., Saigusa, T., Ito, K., Be...
Pp_M_Tokyo_U_N_26h_4	Tero, A., Takagi, S., Saigusa, T., Ito, K., Be...
Pp_M_Tokyo_U_N_26h_5	Tero, A., Takagi, S., Saigusa, T., Ito, K., Be...
...	...
Pv_M_I_U_Hf_18d_3	Rotheray, T.D., Jones, T.H., Fricker, M.D. and...
Pv_M_I_U_Hf_22d_3	Rotheray, T.D., Jones, T.H., Fricker, M.D. and...
Pv_M_I_U_Hf_26d_3	Rotheray, T.D., Jones, T.H., Fricker, M.D. and...
Pv_M_I_U_Hf_38d_3	Rotheray, T.D., Jones, T.H., Fricker, M.D. and...
Pv_M_I_U_Hf_48d_3	Rotheray, T.D., Jones, T.H., Fricker, M.D. and...

[518 rows x 12 columns]

```
[3]: class TreeNode:
      def __init__(self, value):
          self.value = value
          self.left = None
          self.right = None

      color_dict = {
          'A': 'blue',
          'B': 'red',
          'S': 'green',
      }

      node1 = TreeNode('Pp_M_Tokyo_U_N_26h_1.txt')
      color_dict[fnd_df.at[node1.value[:-4], 'substrate']]
```

[3]: 'blue'

```
[4]: #####
      #
      #####
      #####
      #####
```

```

def get_graph_from_file(file, flag):
    if flag:
        print(f'Reading the network file: {file}\n')

    return nx.read_weighted_edgelist(file, nodetype=str)

def graph_properties(G, flag, file):
    '''
    Parameters:
        G (graph)
        flag (int): flag=1 to print graph stats, 0 else
    '''

    n = len(G.nodes())
    m = len(G.edges())
    C = nx.transitivity(G)
    degrees = nx.degree(G)
    kis = [k for _, k in degrees]
    kstd = np.std(kis)
    kmean = np.mean(kis)
    max_edges = n * (n - 1) / 2
    edge_density = m / max_edges
    if nx.is_connected(G):
        average_shortest_path_length = nx.average_shortest_path_length(G,
↪weight='weight')
    else :
        average_shortest_path_length = 0

    betweenness = nx.betweenness_centrality(G, weight='weight')
    non_zero_betweenness = [value for value in betweenness.values() if value !=
↪0]
    avg_betweenness_centrality = np.mean(non_zero_betweenness)

    total_weight = sum(data['weight'] for _, _, data in G.edges(data=True))
    average_edge_weight = total_weight / G.number_of_edges()

    if flag:
        print("Number of nodes (n):", n)
        print("Number of edges (m):", m)
        print("Clustering Coefficient (C):", C)
        print("Mean Degree (kmean):", kmean)
        print("Degree STD (kstd):", kstd)
        print("Edge density (D):", edge_density)
        print("Average shortest path length: ", average_shortest_path_length)
        print("Average betweenness centrality: ", avg_betweenness_centrality)

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    print("Average edge weight: ", average_edge_weight)
    print("\n")

    return [C,kstd,kmean,edge_density,average_shortest_path_length,
            avg_betweenness_centrality,average_edge_weight, file]

def get_all_vectors(directory, flag):
    """
    Parameters:
        directory path (string)
        flag (int): flag=1 to print graph stats, 0 else
    """
    files = os.listdir(directory) # Slicing should be done after calling
    ↪listdir()
    all_graphs = []
    all_graphs_properties = []

    flag_count = 0
    files = list(set(files))

    for file in files:
        file_pathway = directory+file
        if flag_count >= 5:
            flag = 0
            # break

        if os.path.isfile(file_pathway):
            G = get_graph_from_file(file_pathway, flag)
            p = graph_properties(G, flag, file)
            all_graphs.append(G)
            all_graphs_properties.append(p)
            flag_count += 1

    return all_graphs_properties

def euclidean_distance(point1, point2):
    return np.linalg.norm(np.array(point1) - np.array(point2))

def lower_distance_matrix(df):
    """
    Calculate the lower half of the distance matrix for vectors stored in a
    ↪DataFrame using Euclidean

    Args:
        df (pandas.DataFrame): DataFrame containing vectors as rows.

```

```

Returns:
    pandas.DataFrame: Lower half of the distance matrix.
    """

num_rows = df.shape[0]
distance_matrix = np.zeros((num_rows, num_rows))

# Calculate pairwise distances
for i in range(num_rows):
    for j in range(i):
        distance_matrix[i, j] = euclidean_distance(df.iloc[i], df.iloc[j])

# Convert to DataFrame with filename as index
distance_df = pd.DataFrame(distance_matrix, index=df.index, columns=df.
↪index)

return distance_df

```

```

[5]: # A Quick Implementation of UPGMA (Unweighted Pair Group Method with Arithmetic
↪Mean)

# lowest_cell:
# Locates the smallest cell in the table
def lowest_cell(table):
    # Set default to infinity
    min_cell = float("inf")
    x, y = -1, -1

    # Go through every cell, looking for the lowest
    for i in range(len(table)):
        for j in range(len(table[i])):
            if table[i][j] < min_cell:
                min_cell = table[i][j]
                x, y = i, j

    # Return the x, y co-ordinate of cell
    return x, y

# join_labels:
# Combines two labels in a list of labels
def join_labels(labels, a, b):
    # Swap if the indices are not ordered
    if b < a:
        a, b = b, a

```

```

# Join the labels in the first index
labels[a] = [labels[a],labels[b]]

# Remove the (now redundant) label in the second index
del labels[b]

def join_labels(labels, a, b):
    # Swap if the indices are not ordered
    if b < a:
        a, b = b, a

    # Join the labels in the first index
    labels[a] = [labels[a],labels[b]]

    # Remove the (now redundant) label in the second index
    del labels[b]

# join_table:
# Joins the entries of a table on the cell (a, b) by averaging their data
# ↪ entries
def join_table(table, a, b):
    # Swap if the indices are not ordered
    if b < a:
        a, b = b, a

    # For the lower index, reconstruct the entire row (A, i), where i < A
    row = []
    for i in range(0, a):
        row.append((table[a][i] + table[b][i])/2)
    table[a] = row

    # Then, reconstruct the entire column (i, A), where i > A
    # Note: Since the matrix is lower triangular, row b only contains values
    # ↪ for indices < b
    for i in range(a+1, b):
        table[i][a] = (table[i][a]+table[b][i])/2

    # We get the rest of the values from row i
    for i in range(b+1, len(table)):
        table[i][a] = (table[i][a]+table[i][b])/2
        # Remove the (now redundant) second index column entry
        del table[i][b]

    # Remove the (now redundant) second index row
    del table[b]

```

```

# UPGMA:
# Runs the UPGMA algorithm on a labelled table
def UPGMA(table, labels):
    # Until all labels have been joined...
    while len(labels) > 1:
        # Locate lowest cell in the table
        x, y = lowest_cell(table)

        # Join the table on the cell co-ordinates
        join_table(table, x, y)

        # Update the labels accordingly
        join_labels(labels, x, y)

    # Return the final label
    return labels[0]

## A test using an example calculation from http://www.nmsr.org/upgma.htm

# alpha_labels:
# Makes labels from a starting letter to an ending letter
def alpha_labels(start, end):
    labels = []
    for i in range(ord(start), ord(end)+1):
        labels.append(i)
    return labels

def lower_distance_to_M_data(lower_distance_df):
    """
    Convert lower distance DataFrame to M_data format for UPGMA.

    Args:
        lower_distance_df (pandas.DataFrame): Lower half of the distance matrix.

    Returns:
        list: M_data list for UPGMA.
    """
    M_data = []
    num_rows = lower_distance_df.shape[0]
    for i in range(num_rows):
        row_data = []
        for j in range(i):
            row_data.append(lower_distance_df.iloc[i, j])
        M_data.append(row_data)

```

```

    return M_data

class TreeNode:
    def __init__(self, value):
        self.value = value
        self.left = None
        self.right = None

def list_to_binary_tree(lst):
    if isinstance(lst, str):
        return TreeNode(lst)

    root = TreeNode(None)
    root.left = list_to_binary_tree(lst[0])

    if len(lst) == 2:
        root.right = list_to_binary_tree(lst[1])

    return root

def get_node_color(node):
    if node.value is not None:
        print(node.value[:-4])
        return color_dict[fnd_df.at[node.value[:-4], 'substrate']]
    else:
        return 'black'

def plot_tree(node, x, y, spacing, ax=None):
    if ax is None:
        fig, ax = plt.subplots()
    if node is not None:
        ax.plot(x, y, 'o', markersize=12, color='black') # Plot node
        ax.text(x, y, str(node.value), verticalalignment='center',
        ↪horizontalalignment='center') # Add node value as text
        if node.left is not None:
            ax.plot([x, x - spacing], [y, y - 1], '-k') # Plot left edge
            plot_tree(node.left, x - spacing, y - 1, spacing / 2, ax) #
            ↪Recursively plot left subtree
        if node.right is not None:
            ax.plot([x, x + spacing], [y, y - 1], '-k') # Plot right edge
            plot_tree(node.right, x + spacing, y - 1, spacing / 2, ax) #
            ↪Recursively plot right subtree

def replace_with_ln(x):
    if not isinstance(x, str):
        if x > 10:

```



```

        return np.log(x)
    return x
else:
    return x

```

```

[5]: ##### Helpers for clustering #####

# def louvain_clustering(G, flag):
#     partition = community.best_partition(G)
#     num_clusters = len(set(partition.values())) # Count the number of unique
#     ↪ community IDs\

#     if flag:
#         # Draw the graph with nodes colored by community
#         pos = nx.spring_layout(G) # Positions for all nodes
#         plt.figure(figsize=(8, 6))

#         # Draw nodes, coloring them by community
#         node_colors = [partition[n] for n in G.nodes()]
#         nx.draw(G, pos, node_color=node_colors, cmap=plt.cm.tab10,
#         ↪ node_size=20)

#         plt.title("Graph with Nodes Colored by Community (Louvain Method)")
#         plt.show()

#     return partition, num_clusters

```

```

[6]: # G = nx.karate_club_graph

```

```

[7]: # fname1 = 'fungal_networks_data/Conductance_Text_Files/Sc_M_I_U_N_42d_1.txt'
# G = get_graph_from_file(fname1, fname1)

```

```

[8]: ##### TEST HELPER FUNCTIONS FOR ONE FILE #####

# partition, num_clusters = louvain_clustering(G, 1)
# graph_properties(G, 1, fname1)
# print(f'Number of clusters = {num_clusters}')
```

```

[ ]: #####
#####
#####
#####
##### CONDUCTANCE
↪ #####
#####
#####
#####

```

```
#####
```

```
[68]: ##### GET ALL GRAPHS AND PROPERTIES #####
flag = 0
directory = 'fungal_networks_data/Conductance_Text_Files/'
# gmatrix = get_all_vectors(directory, flag)
gmatrix = pd.read_csv('vectors.csv')
column_names = ['C', 'kstd', 'kmean',
    ↪ 'edge_density', 'average_shortest_path_length', 'avg_betweenness centrality', 'average_edge_we
df = pd.DataFrame(gmatrix, columns=column_names)
df.set_index('file', inplace=True)

df
```

```
[68]:
```

	C	kstd	kmean	edge_density \
file				
Rb_M_I_Fc-H_N_20d_1.txt	0.079430	1.627588	2.256983	0.012680
Pv_M_I+4R_U_N_39d_3.txt	0.061126	1.249432	2.233796	0.002588
Rb_M_I_U_N_18d_2.txt	0.069333	1.999792	2.423841	0.008053
Pv_L_I+4xR_U_N_57d_1.txt	0.074269	0.996752	2.641404	0.001855
Pv_M_I_U_N_31d_1.txt	0.041149	1.175326	2.202294	0.001149
...
Pv_M_I_U_Hf_22d_2.txt	0.078322	1.431390	2.581844	0.001847
Pv_M_I+4R_U_N_42d_1.txt	0.052120	1.243313	2.237611	0.001423
Pv_M_I+R_U_N_18d_1.txt	0.115268	1.382307	2.978304	0.005886
Pv_M_I_U_N_18d_3.txt	0.046584	1.176709	2.147410	0.004286
Pv_M_5xI_U_N_22d_3.txt	0.083542	1.122883	2.555283	0.002094

	average_shortest_path_length \
file	
Rb_M_I_Fc-H_N_20d_1.txt	0.000000
Pv_M_I+4R_U_N_39d_3.txt	0.007653
Rb_M_I_U_N_18d_2.txt	22159.387399
Pv_L_I+4xR_U_N_57d_1.txt	0.000000
Pv_M_I_U_N_31d_1.txt	0.016257
...	...
Pv_M_I_U_Hf_22d_2.txt	0.117154
Pv_M_I+4R_U_N_42d_1.txt	0.017854
Pv_M_I+R_U_N_18d_1.txt	0.099048
Pv_M_I_U_N_18d_3.txt	0.007177
Pv_M_5xI_U_N_22d_3.txt	0.000000

	avg_betweenness centrality	average_edge_weight
file		
Rb_M_I_Fc-H_N_20d_1.txt	0.072076	4255.600195
Pv_M_I+4R_U_N_39d_3.txt	0.027278	0.000411
Rb_M_I_U_N_18d_2.txt	0.056738	3140.542753

Pv_L_I+4xR_U_N_57d_1.txt	0.029337	0.006161
Pv_M_I_U_N_31d_1.txt	0.026421	0.000415
...
Pv_M_I_U_Hf_22d_2.txt	0.027449	0.008630
Pv_M_I+4R_U_N_42d_1.txt	0.032957	0.000658
Pv_M_I+R_U_N_18d_1.txt	0.040881	0.017027
Pv_M_I_U_N_18d_3.txt	0.050437	0.000368
Pv_M_5xI_U_N_22d_3.txt	0.030584	0.002475

[270 rows x 7 columns]

```
[11]: df.to_csv('vectors.csv', index=True)
```

```
[75]: df = pd.read_csv('distances.csv')

# Apply the function to all elements of the DataFrame
df_ln = df.applymap(replace_with_ln)
df_ln.to_csv('distances_ln.csv', index=True)
```

```
[69]: ##### GET DISTANCE MATRIX #####
lower_distance_df = lower_distance_matrix(df)
# print(lower_distance_df.to_string(float_format=lambda x: "%.2f" % x))

##### FINISHED #####
```

```
[44]: n = lower_distance_df.shape[0]
symmetric_matrix = np.zeros((n, n))
for i in range(n):
    for j in range(i + 1):
        symmetric_matrix[i, j] = lower_distance_df.iloc[i, j]
        symmetric_matrix[j, i] = lower_distance_df.iloc[i, j]

# Convert symmetric matrix to DataFrame
symmetric_distance_df = pd.DataFrame(symmetric_matrix, index=lower_distance_df.
    ↪index, columns=lower_distance_df.columns)

# Display the symmetric distance DataFrame
symmetric_distance_df
```

```
[44]: file Rb_M_I_Fc-H_N_20d_1.txt Pv_M_I+4R_U_N_39d_3.txt \
file
Rb_M_I_Fc-H_N_20d_1.txt 0.000000 4255.599801
Pv_M_I+4R_U_N_39d_3.txt 4255.599801 0.000000
Rb_M_I_U_N_18d_2.txt 22187.424438 22380.820306
Pv_L_I+4xR_U_N_57d_1.txt 4255.594098 0.479854
Pv_M_I_U_N_31d_1.txt 4255.599805 0.083426
...
```

Pv_M_I_U_Hf_22d_2.txt	4255.591584	0.408167
Pv_M_I+4R_U_N_42d_1.txt	4255.599555	0.016457
Pv_M_I+R_U_N_18d_1.txt	4255.583238	0.764005
Pv_M_I_U_N_18d_3.txt	4255.599853	0.116198
Pv_M_5xI_U_N_22d_3.txt	4255.597761	0.346330

file	Rb_M_I_U_N_18d_2.txt	Pv_L_I+4xR_U_N_57d_1.txt	\
file			

Rb_M_I_Fc-H_N_20d_1.txt	22187.424438	4255.594098
Pv_M_I+4R_U_N_39d_3.txt	22380.820306	0.479854
Rb_M_I_U_N_18d_2.txt	0.000000	22380.827086
Pv_L_I+4xR_U_N_57d_1.txt	22380.827086	0.000000
Pv_M_I_U_N_31d_1.txt	22380.811789	0.475510

...
Pv_M_I_U_Hf_22d_2.txt	22380.710729	0.454102
Pv_M_I+4R_U_N_42d_1.txt	22380.810171	0.474019
Pv_M_I+R_U_N_18d_1.txt	22380.727485	0.523367
Pv_M_I_U_N_18d_3.txt	22380.820786	0.526988
Pv_M_5xI_U_N_22d_3.txt	22380.827597	0.153059

file	Pv_M_I_U_N_31d_1.txt	Rb_M_I_U_N_26d_3.txt	\
file			

Rb_M_I_Fc-H_N_20d_1.txt	4255.599805	13109.480470
Pv_M_I+4R_U_N_39d_3.txt	0.083426	13047.500922
Rb_M_I_U_N_18d_2.txt	22380.811789	9334.383427
Pv_L_I+4xR_U_N_57d_1.txt	0.475510	13047.507640
Pv_M_I_U_N_31d_1.txt	0.000000	13047.492416

...
Pv_M_I_U_Hf_22d_2.txt	0.470381	13047.391406
Pv_M_I+4R_U_N_42d_1.txt	0.077687	13047.490797
Pv_M_I+R_U_N_18d_1.txt	0.811103	13047.408071
Pv_M_I_U_N_18d_3.txt	0.060932	13047.501404
Pv_M_5xI_U_N_22d_3.txt	0.359771	13047.508182

file	Pp_M_Tokyo_U_N_26h_1.txt	Pv_M_I+4R_U_N_36d_2.txt	\
file			

Rb_M_I_Fc-H_N_20d_1.txt	77439.075233	4255.596544
Pv_M_I+4R_U_N_39d_3.txt	77899.735886	0.152118
Rb_M_I_U_N_18d_2.txt	55519.328401	22380.772713
Pv_L_I+4xR_U_N_57d_1.txt	77899.742686	0.344897
Pv_M_I_U_N_31d_1.txt	77899.727361	0.175327

...
Pv_M_I_U_Hf_22d_2.txt	77899.626277	0.320894
Pv_M_I+4R_U_N_42d_1.txt	77899.725745	0.144057
Pv_M_I+R_U_N_18d_1.txt	77899.643080	0.639594
Pv_M_I_U_N_18d_3.txt	77899.736364	0.228910
Pv_M_5xI_U_N_22d_3.txt	77899.743186	0.210226

file	Pv_M_I_U_N_27d_1.txt	Pv_L_I+4xR_U_N_36d_4.txt	...	\
file			...	
Rb_M_I_Fc-H_N_20d_1.txt	4255.599854	4255.593781	...	
Pv_M_I+4R_U_N_39d_3.txt	0.094394	0.283851	...	
Rb_M_I_U_N_18d_2.txt	22380.813432	22380.827038	...	
Pv_L_I+4xR_U_N_57d_1.txt	0.484274	0.216259	...	
Pv_M_I_U_N_31d_1.txt	0.014589	0.304099	...	
...	
Pv_M_I_U_Hf_22d_2.txt	0.483552	0.305944	...	
Pv_M_I+4R_U_N_42d_1.txt	0.089558	0.279597	...	
Pv_M_I+R_U_N_18d_1.txt	0.824218	0.535658	...	
Pv_M_I_U_N_18d_3.txt	0.054333	0.357177	...	
Pv_M_5xI_U_N_22d_3.txt	0.370741	0.066788	...	

file	Pp_M_Tokyo_U_N_26h_11.txt	Pv_M_5xI_U_N_33d_3.txt	...	\
file			...	
Rb_M_I_Fc-H_N_20d_1.txt	4255.572255	4255.596232	...	
Pv_M_I+4R_U_N_39d_3.txt	0.922752	0.456321	...	
Rb_M_I_U_N_18d_2.txt	22380.592636	22380.786547	...	
Pv_L_I+4xR_U_N_57d_1.txt	0.490105	0.156095	...	
Pv_M_I_U_N_31d_1.txt	0.926128	0.477296	...	
...	
Pv_M_I_U_Hf_22d_2.txt	0.724592	0.313506	...	
Pv_M_I+4R_U_N_42d_1.txt	0.915231	0.451338	...	
Pv_M_I+R_U_N_18d_1.txt	0.510405	0.392255	...	
Pv_M_I_U_N_18d_3.txt	0.977301	0.532227	...	
Pv_M_5xI_U_N_22d_3.txt	0.597656	0.128384	...	

file	Pv_M_I_U_Hf_26d_3.txt	Pv_M_I_Fc_Hf_48d_2.txt	...	\
file			...	
Rb_M_I_Fc-H_N_20d_1.txt	4255.589776	4255.583939	...	
Pv_M_I+4R_U_N_39d_3.txt	0.591240	0.450383	...	
Rb_M_I_U_N_18d_2.txt	22380.690440	22380.825651	...	
Pv_L_I+4xR_U_N_57d_1.txt	0.542775	0.443469	...	
Pv_M_I_U_N_31d_1.txt	0.653139	0.514954	...	
...	
Pv_M_I_U_Hf_22d_2.txt	0.184921	0.132814	...	
Pv_M_I+4R_U_N_42d_1.txt	0.589131	0.450742	...	
Pv_M_I+R_U_N_18d_1.txt	0.269067	0.357774	...	
Pv_M_I_U_N_18d_3.txt	0.700698	0.561509	...	
Pv_M_5xI_U_N_22d_3.txt	0.454164	0.328361	...	

file	Pv_M_I+4R_U_N_27d_1.txt	Pv_M_I_U_Hf_22d_2.txt	...	\
file			...	
Rb_M_I_Fc-H_N_20d_1.txt	4255.599709	4255.591584	...	
Pv_M_I+4R_U_N_39d_3.txt	0.043477	0.408167	...	

Rb_M_I_U_N_18d_2.txt	22380.815728	22380.710729
Pv_L_I+4xR_U_N_57d_1.txt	0.450236	0.454102
Pv_M_I_U_N_31d_1.txt	0.112891	0.470381
...
Pv_M_I_U_Hf_22d_2.txt	0.367124	0.000000
Pv_M_I+4R_U_N_42d_1.txt	0.042652	0.405599
Pv_M_I+R_U_N_18d_1.txt	0.720917	0.401932
Pv_M_I_U_N_18d_3.txt	0.154849	0.517011
Pv_M_5xI_U_N_22d_3.txt	0.311935	0.331183
file	Pv_M_I+4R_U_N_42d_1.txt	Pv_M_I+R_U_N_18d_1.txt \
file		
Rb_M_I_Fc-H_N_20d_1.txt	4255.599555	4255.583238
Pv_M_I+4R_U_N_39d_3.txt	0.016457	0.764005
Rb_M_I_U_N_18d_2.txt	22380.810171	22380.727485
Pv_L_I+4xR_U_N_57d_1.txt	0.474019	0.523367
Pv_M_I_U_N_31d_1.txt	0.077687	0.811103
...
Pv_M_I_U_Hf_22d_2.txt	0.405599	0.401932
Pv_M_I+4R_U_N_42d_1.txt	0.000000	0.760839
Pv_M_I+R_U_N_18d_1.txt	0.760839	0.000000
Pv_M_I_U_N_18d_3.txt	0.114152	0.863819
Pv_M_5xI_U_N_22d_3.txt	0.341664	0.507343
file	Pv_M_I_U_N_18d_3.txt	Pv_M_5xI_U_N_22d_3.txt
file		
Rb_M_I_Fc-H_N_20d_1.txt	4255.599853	4255.597761
Pv_M_I+4R_U_N_39d_3.txt	0.116198	0.346330
Rb_M_I_U_N_18d_2.txt	22380.820786	22380.827597
Pv_L_I+4xR_U_N_57d_1.txt	0.526988	0.153059
Pv_M_I_U_N_31d_1.txt	0.060932	0.359771
...
Pv_M_I_U_Hf_22d_2.txt	0.517011	0.331183
Pv_M_I+4R_U_N_42d_1.txt	0.114152	0.341664
Pv_M_I+R_U_N_18d_1.txt	0.863819	0.507343
Pv_M_I_U_N_18d_3.txt	0.000000	0.413615
Pv_M_5xI_U_N_22d_3.txt	0.413615	0.000000

[270 rows x 270 columns]

```
[59]: def func(row):
        index_name = row.name
        return fnd_df.at[index_name[:-4], 'substrate']

# Apply the function to create a new column
symmetric_distance_df['substrate'] = symmetric_distance_df.apply(func, axis=1)
```

```
symmetric_distance_df.to_csv('distances_substrate.csv', index_label='index',
    ↪header=True)
```

```
[50]: # np.savetxt('distances.csv', symmetric_distance_df, delimiter=',')
# # Assuming df is your DataFrame
# symmetric_distance_df.to_csv('distances.csv', index_label='index',
    ↪header=True)
```

```
[48]: import scipy.io
import numpy as np

# Save the distance matrix and row/column names to a .mat file
scipy.io.savemat('distance_matrix.mat', {'distances': symmetric_distance_df,
    ↪'row_names': symmetric_distance_df.index.tolist(),
    ↪'column_names': symmetric_distance_df.
    ↪columns.tolist()})
```

```
[13]: #####          SMALL TEST ON THING          #####

M_labels = lower_distance_df.index.tolist()
M = lower_distance_to_M_data(lower_distance_df)
lst = UPGMA(M, M_labels)
```

```
[65]: symmetric_distance_df = symmetric_distance_df.drop(columns=['substrate'])
```

```
[29]: from scipy.cluster.hierarchy import linkage, dendrogram
import matplotlib.pyplot as plt
from scipy.spatial.distance import squareform

# Your distance matrix
distances = pd.read_csv('distances_ln.csv')
# # distances = distances.drop(df.columns[[1]], axis=1)
dist_sub = distances

dist_sub.set_index('index', inplace=True)

unnamed_columns = [col for col in dist_sub.columns if col.startswith('Unnamed')]
dist_sub = dist_sub.drop(columns=unnamed_columns)

dist_sub
```

```
[29]:                                     Rb_M_I_Fc-H_N_20d_1.txt  Pv_M_I+4R_U_N_39d_3.txt  \
index
Rb_M_I_Fc-H_N_20d_1.txt                0.000000                8.355991
Pv_M_I+4R_U_N_39d_3.txt                8.355991                0.000000
Rb_M_I_U_N_18d_2.txt                  10.007281                10.015960
```

Pv_L_I+4xR_U_N_57d_1.txt	8.355990	0.479854
Pv_M_I_U_N_31d_1.txt	8.355991	0.083426
...
Pv_M_I_U_Hf_22d_2.txt	8.355989	0.408167
Pv_M_I+4R_U_N_42d_1.txt	8.355991	0.016457
Pv_M_I+R_U_N_18d_1.txt	8.355987	0.764005
Pv_M_I_U_N_18d_3.txt	8.355991	0.116198
Pv_M_5xI_U_N_22d_3.txt	8.355991	0.346330

	Rb_M_I_U_N_18d_2.txt	Pv_L_I+4xR_U_N_57d_1.txt \
index		
Rb_M_I_Fc-H_N_20d_1.txt	10.007281	8.355990
Pv_M_I+4R_U_N_39d_3.txt	10.015960	0.479854
Rb_M_I_U_N_18d_2.txt	0.000000	10.015960
Pv_L_I+4xR_U_N_57d_1.txt	10.015960	0.000000
Pv_M_I_U_N_31d_1.txt	10.015959	0.475510
...
Pv_M_I_U_Hf_22d_2.txt	10.015955	0.454102
Pv_M_I+4R_U_N_42d_1.txt	10.015959	0.474019
Pv_M_I+R_U_N_18d_1.txt	10.015955	0.523367
Pv_M_I_U_N_18d_3.txt	10.015960	0.526988
Pv_M_5xI_U_N_22d_3.txt	10.015960	0.153059

	Pv_M_I_U_N_31d_1.txt	Rb_M_I_U_N_26d_3.txt \
index		
Rb_M_I_Fc-H_N_20d_1.txt	8.355991	9.481091
Pv_M_I+4R_U_N_39d_3.txt	0.083426	9.476352
Rb_M_I_U_N_18d_2.txt	10.015959	9.141460
Pv_L_I+4xR_U_N_57d_1.txt	0.475510	9.476352
Pv_M_I_U_N_31d_1.txt	0.000000	9.476351
...
Pv_M_I_U_Hf_22d_2.txt	0.470381	9.476344
Pv_M_I+4R_U_N_42d_1.txt	0.077687	9.476351
Pv_M_I+R_U_N_18d_1.txt	0.811103	9.476345
Pv_M_I_U_N_18d_3.txt	0.060932	9.476352
Pv_M_5xI_U_N_22d_3.txt	0.359771	9.476352

	Pp_M_Tokyo_U_N_26h_1.txt	Pv_M_I+4R_U_N_36d_2.txt \
index		
Rb_M_I_Fc-H_N_20d_1.txt	11.257247	8.355990
Pv_M_I+4R_U_N_39d_3.txt	11.263178	0.152118
Rb_M_I_U_N_18d_2.txt	10.924486	10.015958
Pv_L_I+4xR_U_N_57d_1.txt	11.263178	0.344897
Pv_M_I_U_N_31d_1.txt	11.263178	0.175327
...
Pv_M_I_U_Hf_22d_2.txt	11.263176	0.320894
Pv_M_I+4R_U_N_42d_1.txt	11.263178	0.144057

Pv_M_I+R_U_N_18d_1.txt	11.263177	0.639594
Pv_M_I_U_N_18d_3.txt	11.263178	0.228910
Pv_M_5xI_U_N_22d_3.txt	11.263178	0.210226

	Pv_M_I_U_N_27d_1.txt	Pv_L_I+4xR_U_N_36d_4.txt	...	\
index				
Rb_M_I_Fc-H_N_20d_1.txt	8.355991	8.355990	...	
Pv_M_I+4R_U_N_39d_3.txt	0.094394	0.283851	...	
Rb_M_I_U_N_18d_2.txt	10.015959	10.015960	...	
Pv_L_I+4xR_U_N_57d_1.txt	0.484274	0.216259	...	
Pv_M_I_U_N_31d_1.txt	0.014589	0.304099	...	
...	
Pv_M_I_U_Hf_22d_2.txt	0.483552	0.305944	...	
Pv_M_I+4R_U_N_42d_1.txt	0.089558	0.279597	...	
Pv_M_I+R_U_N_18d_1.txt	0.824218	0.535658	...	
Pv_M_I_U_N_18d_3.txt	0.054333	0.357177	...	
Pv_M_5xI_U_N_22d_3.txt	0.370741	0.066788	...	

	Pp_M_Tokyo_U_N_26h_11.txt	Pv_M_5xI_U_N_33d_3.txt	...	\
index				
Rb_M_I_Fc-H_N_20d_1.txt	8.355985	8.355990		
Pv_M_I+4R_U_N_39d_3.txt	0.922752	0.456321		
Rb_M_I_U_N_18d_2.txt	10.015949	10.015958		
Pv_L_I+4xR_U_N_57d_1.txt	0.490105	0.156095		
Pv_M_I_U_N_31d_1.txt	0.926128	0.477296		
...		
Pv_M_I_U_Hf_22d_2.txt	0.724592	0.313506		
Pv_M_I+4R_U_N_42d_1.txt	0.915231	0.451338		
Pv_M_I+R_U_N_18d_1.txt	0.510405	0.392255		
Pv_M_I_U_N_18d_3.txt	0.977301	0.532227		
Pv_M_5xI_U_N_22d_3.txt	0.597656	0.128384		

	Pv_M_I_U_Hf_26d_3.txt	Pv_M_I_Fc_Hf_48d_2.txt	...	\
index				
Rb_M_I_Fc-H_N_20d_1.txt	8.355989	8.355987		
Pv_M_I+4R_U_N_39d_3.txt	0.591240	0.450383		
Rb_M_I_U_N_18d_2.txt	10.015954	10.015960		
Pv_L_I+4xR_U_N_57d_1.txt	0.542775	0.443469		
Pv_M_I_U_N_31d_1.txt	0.653139	0.514954		
...		
Pv_M_I_U_Hf_22d_2.txt	0.184921	0.132814		
Pv_M_I+4R_U_N_42d_1.txt	0.589131	0.450742		
Pv_M_I+R_U_N_18d_1.txt	0.269067	0.357774		
Pv_M_I_U_N_18d_3.txt	0.700698	0.561509		
Pv_M_5xI_U_N_22d_3.txt	0.454164	0.328361		

	Pv_M_I+4R_U_N_27d_1.txt	Pv_M_I_U_Hf_22d_2.txt	...	\
--	-------------------------	-----------------------	-----	---

index		
Rb_M_I_Fc-H_N_20d_1.txt	8.355991	8.355989
Pv_M_I+4R_U_N_39d_3.txt	0.043477	0.408167
Rb_M_I_U_N_18d_2.txt	10.015959	10.015955
Pv_L_I+4xR_U_N_57d_1.txt	0.450236	0.454102
Pv_M_I_U_N_31d_1.txt	0.112891	0.470381
...
Pv_M_I_U_Hf_22d_2.txt	0.367124	0.000000
Pv_M_I+4R_U_N_42d_1.txt	0.042652	0.405599
Pv_M_I+R_U_N_18d_1.txt	0.720917	0.401932
Pv_M_I_U_N_18d_3.txt	0.154849	0.517011
Pv_M_5xI_U_N_22d_3.txt	0.311935	0.331183

	Pv_M_I+4R_U_N_42d_1.txt	Pv_M_I+R_U_N_18d_1.txt	\
--	-------------------------	------------------------	---

index		
Rb_M_I_Fc-H_N_20d_1.txt	8.355991	8.355987
Pv_M_I+4R_U_N_39d_3.txt	0.016457	0.764005
Rb_M_I_U_N_18d_2.txt	10.015959	10.015955
Pv_L_I+4xR_U_N_57d_1.txt	0.474019	0.523367
Pv_M_I_U_N_31d_1.txt	0.077687	0.811103
...
Pv_M_I_U_Hf_22d_2.txt	0.405599	0.401932
Pv_M_I+4R_U_N_42d_1.txt	0.000000	0.760839
Pv_M_I+R_U_N_18d_1.txt	0.760839	0.000000
Pv_M_I_U_N_18d_3.txt	0.114152	0.863819
Pv_M_5xI_U_N_22d_3.txt	0.341664	0.507343

	Pv_M_I_U_N_18d_3.txt	Pv_M_5xI_U_N_22d_3.txt
--	----------------------	------------------------

index		
Rb_M_I_Fc-H_N_20d_1.txt	8.355991	8.355991
Pv_M_I+4R_U_N_39d_3.txt	0.116198	0.346330
Rb_M_I_U_N_18d_2.txt	10.015960	10.015960
Pv_L_I+4xR_U_N_57d_1.txt	0.526988	0.153059
Pv_M_I_U_N_31d_1.txt	0.060932	0.359771
...
Pv_M_I_U_Hf_22d_2.txt	0.517011	0.331183
Pv_M_I+4R_U_N_42d_1.txt	0.114152	0.341664
Pv_M_I+R_U_N_18d_1.txt	0.863819	0.507343
Pv_M_I_U_N_18d_3.txt	0.000000	0.413615
Pv_M_5xI_U_N_22d_3.txt	0.413615	0.000000

[270 rows x 270 columns]

```
[116]: # graph
import matplotlib.pyplot as plt
from scipy.cluster.hierarchy import dendrogram, linkage
```

```

color_mapping = {
    'Pp': 0,
    'Pv': 1,
    'Ag': 2,
    'Pi': 3,
    'Rb': 4,
    'Sc': 5,
    'none': 6
}

def get_leaf_color(filename):
    if filename[-4:] == '.txt':
        value = filename[0:2]
        return color_mapping[value]
    else:
        return color_mapping['none']

# Assuming 'df' contains attributes for each leaf node
# Pass the leaf attributes to the color calculation function
leaf_colors = [get_leaf_color(index_name) for index_name in dist_sub.index]

# Your distance matrix
distances = pd.read_csv('conductance_distances_ln.csv')
distances.set_index('index', inplace=True)
distances = distances.drop(distances.columns[[0]], axis=1)

tree = linkage(distances, method='average', optimal_ordering=False) # Use
    ↳ 'average' method for UPGMA

# Plot dendrogram
plt.figure(figsize=(20, 20))
# dendrogram(tree, labels=distances.index, leaf_rotation=90)
dendrogram(tree, labels=distances.index, leaf_rotation=0, orientation="left",
    ↳ color_threshold=240, above_threshold_color='grey')
plt.title('UPGMA Dendrogram Using Conductance')
plt.xlabel('Data Points')
plt.ylabel('Distance')
# plt.show()

my_palette = plt.cm.get_cmap('RdYlGn', 7)

ax = plt.gca()
xlbls = ax.get_ymajorticklabels()
num=-1
for lbl in xlbls:

```

```

        val = get_leaf_color(lbl.get_text())
        lbl.set_color(my_palette(val))

# Show the graph
plt.show()

# Define labels for each color
color_labels = ['Pp', 'Pv', 'Ag', 'Pi', 'Rb', 'Sc']

# Create a dictionary to map colors to labels
color_map = {my_palette(i): label for i, label in enumerate(color_labels)}

# Display the colors and their corresponding labels
for color, label in color_map.items():
    print(f'{label}: {color}')

# Plot example colors
plt.figure(figsize=(6, 1))
for color, label in color_map.items():
    plt.plot([0, 1], [0, 0], color=color, label=label, linewidth=4)
plt.xlim(0, 1)
plt.ylim(-0.5, 0.5)
plt.yticks([])
plt.legend(loc='center', ncol=len(color_labels))
plt.title('Color Key')
plt.show()

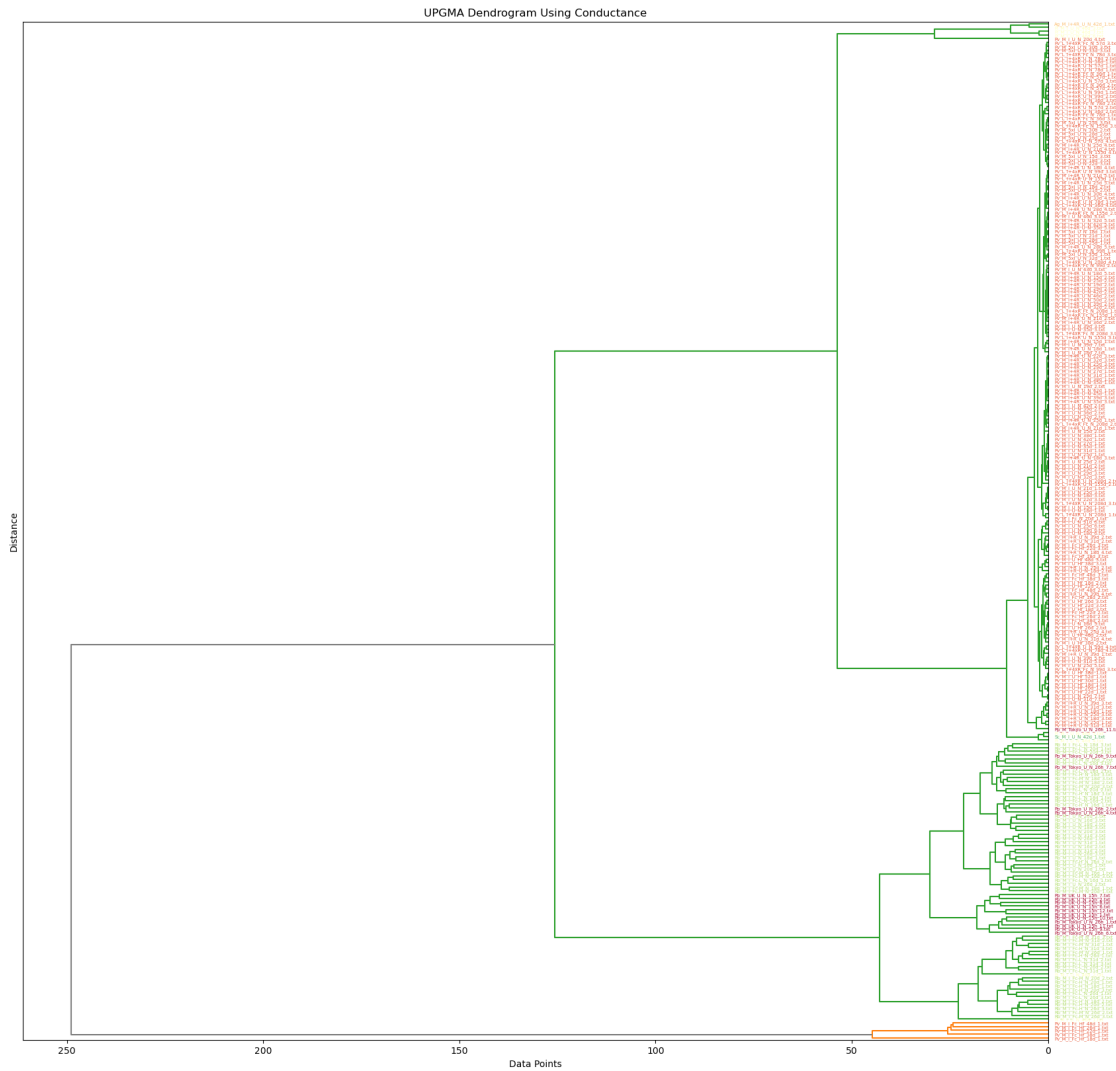
```

/tmp/ipykernel_375/3232813409.py:32: ClusterWarning: scipy.cluster: The symmetric non-negative hollow observation matrix looks suspiciously like an uncondensed distance matrix

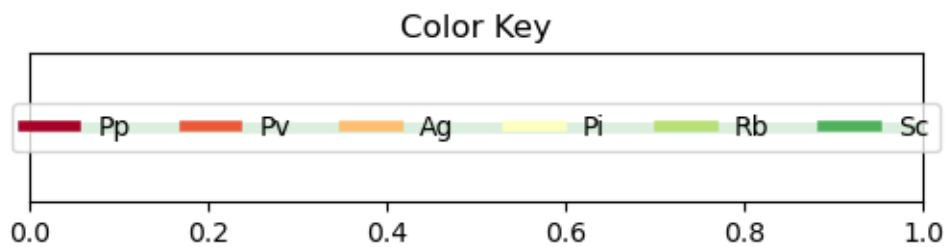
```

    tree = linkage(distances, method='average', optimal_ordering=False) # Use
'average' method for UPGMA

```



Pp: (0.6470588235294118, 0.0, 0.14901960784313725, 1.0)
Pv: (0.918954248366013, 0.3477124183006535, 0.22614379084967318, 1.0)
Ag: (0.9934640522875817, 0.7477124183006535, 0.4352941176470587, 1.0)
Pi: (1.0, 1.0, 0.7490196078431373, 1.0)
Rb: (0.7176470588235295, 0.8797385620915033, 0.45882352941176474, 1.0)
Sc: (0.3006535947712421, 0.6928104575163401, 0.3633986928104576, 1.0)




```

...
Pv_M_5xI_U_N_21d_2.txt          0.0
Pv_M_I_U_N_18d_6.txt            0.0
Pv_L_I+4xR_U_N_78d_1.txt        0.0
Pv_M_I_U_N_32d_3.txt            0.0
Pv_M_I+4R_U_N_31d_1.txt         0.0

                                avg_betweenness centrality  average edge weight
file
Pv_M_I+4R_U_N_27d_1.txt          0.022698                0.009950
Rb_M_I_Fc-L_N_16d_3.txt          0.053408                0.020824
Pv_M_I+4R_U_N_29d_3.txt          0.035083                0.019230
Pv_M_I_Fc_Hf_38d_2.txt          0.019169                0.003332
Pv_M_I_U_N_15d_1.txt            0.001925                0.027958
...
Pv_M_5xI_U_N_21d_2.txt          0.029372                0.008983
Pv_M_I_U_N_18d_6.txt            0.037684                0.025245
Pv_L_I+4xR_U_N_78d_1.txt        0.029216                0.005701
Pv_M_I_U_N_32d_3.txt            0.031155                0.010983
Pv_M_I+4R_U_N_31d_1.txt         0.019604                0.009367

[270 rows x 7 columns]

```

```
[80]: # df.to_csv('path_score_vectors.csv', index=True)
```

```
[88]: ps_lower_dist = lower_distance_matrix(psvecs)
# ps_lower_dist
```

```
[103]: ##### MAKE SYMETRICAL #####
n = ps_lower_dist.shape[0]
ps_symetric_dist = np.zeros((n, n))
for i in range(n):
    for j in range(i + 1):
        ps_symetric_dist[i, j] = ps_lower_dist.iloc[i, j]
        ps_symetric_dist[j, i] = ps_lower_dist.iloc[i, j]

# Convert symmetric matrix to DataFrame
ps_symetric_dist = pd.DataFrame(ps_symetric_dist, index=ps_lower_dist.index,
                                columns=ps_lower_dist.columns)

# Display the symmetric distance DataFrame
ps_symetric_dist
```

```
[103]: file          Pv_M_I+4R_U_N_27d_1.txt  Rb_M_I_Fc-L_N_16d_3.txt  \
file
Pv_M_I+4R_U_N_27d_1.txt          0.000000                0.825634
Rb_M_I_Fc-L_N_16d_3.txt          0.825634                0.000000
```

Pv_M_I+4R_U_N_29d_3.txt	0.186360	0.999453
Pv_M_I_Fc_Hf_38d_2.txt	0.285853	0.556588
Pv_M_I_U_N_15d_1.txt	0.550161	1.329045
...
Pv_M_5xI_U_N_21d_2.txt	0.105457	0.927568
Pv_M_I_U_N_18d_6.txt	0.436877	0.639594
Pv_L_I+4xR_U_N_78d_1.txt	0.263120	1.066627
Pv_M_I_U_N_32d_3.txt	0.230314	1.048077
Pv_M_I+4R_U_N_31d_1.txt	0.042060	0.854447

file Pv_M_I+4R_U_N_29d_3.txt Pv_M_I_Fc_Hf_38d_2.txt \

file

Pv_M_I+4R_U_N_27d_1.txt	0.186360	0.285853
Rb_M_I_Fc-L_N_16d_3.txt	0.999453	0.556588
Pv_M_I+4R_U_N_29d_3.txt	0.000000	0.470338
Pv_M_I_Fc_Hf_38d_2.txt	0.470338	0.000000
Pv_M_I_U_N_15d_1.txt	0.377332	0.828894
...
Pv_M_5xI_U_N_21d_2.txt	0.111899	0.385329
Pv_M_I_U_N_18d_6.txt	0.532882	0.409936
Pv_L_I+4xR_U_N_78d_1.txt	0.199583	0.513373
Pv_M_I_U_N_32d_3.txt	0.072756	0.515295
Pv_M_I+4R_U_N_31d_1.txt	0.151238	0.321974

file Pv_M_I_U_N_15d_1.txt Pv_M_I_Fc_Hf_22d_1.txt \

file

Pv_M_I+4R_U_N_27d_1.txt	0.550161	0.035963
Rb_M_I_Fc-L_N_16d_3.txt	1.329045	0.848524
Pv_M_I+4R_U_N_29d_3.txt	0.377332	0.177133
Pv_M_I_Fc_Hf_38d_2.txt	0.828894	0.302399
Pv_M_I_U_N_15d_1.txt	0.000000	0.545517
...
Pv_M_5xI_U_N_21d_2.txt	0.465073	0.087499
Pv_M_I_U_N_18d_6.txt	0.757071	0.471237
Pv_L_I+4xR_U_N_78d_1.txt	0.451924	0.232973
Pv_M_I_U_N_32d_3.txt	0.333874	0.217843
Pv_M_I+4R_U_N_31d_1.txt	0.510708	0.058807

file Pv_M_I_Fc_Hf_22d_3.txt Pv_M_I+R_U_N_25d_2.txt \

file

Pv_M_I+4R_U_N_27d_1.txt	0.510857	0.683918
Rb_M_I_Fc-L_N_16d_3.txt	0.450957	0.205435
Pv_M_I+4R_U_N_29d_3.txt	0.650687	0.847215
Pv_M_I_Fc_Hf_38d_2.txt	0.362617	0.446525
Pv_M_I_U_N_15d_1.txt	0.923710	1.155070
...
Pv_M_5xI_U_N_21d_2.txt	0.600694	0.784287

Pv_M_I_U_N_18d_6.txt	0.192404	0.442198
Pv_L_I+4xR_U_N_78d_1.txt	0.764886	0.937136
Pv_M_I_U_N_32d_3.txt	0.693097	0.894882
Pv_M_I+4R_U_N_31d_1.txt	0.522942	0.706595
file	Pv_L_I+4xR_U_N_155d_1.txt	Rb_M_I_Fc-H_N_16d_3.txt \
file		
Pv_M_I+4R_U_N_27d_1.txt	0.224900	0.876111
Rb_M_I_Fc-L_N_16d_3.txt	1.044908	0.278068
Pv_M_I+4R_U_N_29d_3.txt	0.120490	1.026358
Pv_M_I_Fc_Hf_38d_2.txt	0.499113	0.659316
Pv_M_I_U_N_15d_1.txt	0.396883	1.304863
...
Pv_M_5xI_U_N_21d_2.txt	0.121736	0.972901
Pv_M_I_U_N_18d_6.txt	0.612275	0.553889
Pv_L_I+4xR_U_N_78d_1.txt	0.087742	1.132182
Pv_M_I_U_N_32d_3.txt	0.088713	1.073162
Pv_M_I+4R_U_N_31d_1.txt	0.204685	0.894013
file	...	Pv_L_I+4xR_Fc_N_99d_3.txt \
file	...	
Pv_M_I+4R_U_N_27d_1.txt	...	0.166148
Rb_M_I_Fc-L_N_16d_3.txt	...	0.892280
Pv_M_I+4R_U_N_29d_3.txt	...	0.252102
Pv_M_I_Fc_Hf_38d_2.txt	...	0.338369
Pv_M_I_U_N_15d_1.txt	...	0.606389
...
Pv_M_5xI_U_N_21d_2.txt	...	0.166587
Pv_M_I_U_N_18d_6.txt	...	0.593193
Pv_L_I+4xR_U_N_78d_1.txt	...	0.197739
Pv_M_I_U_N_32d_3.txt	...	0.276801
Pv_M_I+4R_U_N_31d_1.txt	...	0.190154
file	Rb_M_I_Fc-H_N_20d_1.txt	Rb_M_I_Fc-M_N_31d_2.txt \
file		
Pv_M_I+4R_U_N_27d_1.txt	0.855142	0.618249
Rb_M_I_Fc-L_N_16d_3.txt	0.428068	1.351946
Pv_M_I+4R_U_N_29d_3.txt	0.992467	0.451047
Pv_M_I_Fc_Hf_38d_2.txt	0.669687	0.875011
Pv_M_I_U_N_15d_1.txt	1.257518	0.329602
...
Pv_M_5xI_U_N_21d_2.txt	0.943964	0.561759
Pv_M_I_U_N_18d_6.txt	0.562132	0.825626
Pv_L_I+4xR_U_N_78d_1.txt	1.096581	0.571152
Pv_M_I_U_N_32d_3.txt	1.036648	0.456839
Pv_M_I+4R_U_N_31d_1.txt	0.870829	0.582019

file	Pv_M_I+4R_U_N_15d_1.txt	Rb_M_I_Fc-L_N_31d_1.txt	\
file			
Pv_M_I+4R_U_N_27d_1.txt	0.079561	1.455208	
Rb_M_I_Fc-L_N_16d_3.txt	0.797104	0.894267	
Pv_M_I+4R_U_N_29d_3.txt	0.246994	1.568770	
Pv_M_I_Fc_Hf_38d_2.txt	0.244624	1.288085	
Pv_M_I_U_N_15d_1.txt	0.615036	1.767391	
...	
Pv_M_5xI_U_N_21d_2.txt	0.152573	1.542095	
Pv_M_I_U_N_18d_6.txt	0.475215	1.051327	
Pv_L_I+4xR_U_N_78d_1.txt	0.272600	1.711029	
Pv_M_I_U_N_32d_3.txt	0.285785	1.614433	
Pv_M_I+4R_U_N_31d_1.txt	0.120149	1.462166	
file	Pv_M_5xI_U_N_21d_2.txt	Pv_M_I_U_N_18d_6.txt	\
file			
Pv_M_I+4R_U_N_27d_1.txt	0.105457	0.436877	
Rb_M_I_Fc-L_N_16d_3.txt	0.927568	0.639594	
Pv_M_I+4R_U_N_29d_3.txt	0.111899	0.532882	
Pv_M_I_Fc_Hf_38d_2.txt	0.385329	0.409936	
Pv_M_I_U_N_15d_1.txt	0.465073	0.757071	
...	
Pv_M_5xI_U_N_21d_2.txt	0.000000	0.506170	
Pv_M_I_U_N_18d_6.txt	0.506170	0.000000	
Pv_L_I+4xR_U_N_78d_1.txt	0.171419	0.673984	
Pv_M_I_U_N_32d_3.txt	0.134457	0.569607	
Pv_M_I+4R_U_N_31d_1.txt	0.086053	0.434129	
file	Pv_L_I+4xR_U_N_78d_1.txt	Pv_M_I_U_N_32d_3.txt	\
file			
Pv_M_I+4R_U_N_27d_1.txt	0.263120	0.230314	
Rb_M_I_Fc-L_N_16d_3.txt	1.066627	1.048077	
Pv_M_I+4R_U_N_29d_3.txt	0.199583	0.072756	
Pv_M_I_Fc_Hf_38d_2.txt	0.513373	0.515295	
Pv_M_I_U_N_15d_1.txt	0.451924	0.333874	
...	
Pv_M_5xI_U_N_21d_2.txt	0.171419	0.134457	
Pv_M_I_U_N_18d_6.txt	0.673984	0.569607	
Pv_L_I+4xR_U_N_78d_1.txt	0.000000	0.170706	
Pv_M_I_U_N_32d_3.txt	0.170706	0.000000	
Pv_M_I+4R_U_N_31d_1.txt	0.254308	0.196891	
file	Pv_M_I+4R_U_N_31d_1.txt		
file			
Pv_M_I+4R_U_N_27d_1.txt	0.042060		
Rb_M_I_Fc-L_N_16d_3.txt	0.854447		
Pv_M_I+4R_U_N_29d_3.txt	0.151238		

Pv_M_I_Fc_Hf_38d_2.txt	0.321974
Pv_M_I_U_N_15d_1.txt	0.510708
...	...
Pv_M_5xI_U_N_21d_2.txt	0.086053
Pv_M_I_U_N_18d_6.txt	0.434129
Pv_L_I+4xR_U_N_78d_1.txt	0.254308
Pv_M_I_U_N_32d_3.txt	0.196891
Pv_M_I+4R_U_N_31d_1.txt	0.000000

[270 rows x 270 columns]

```
[104]: ps_symetric_dist_2 = ps_symetric_dist[~ps_symetric_dist.index.duplicated()]
ps_symetric_dist_2
```

```
[104]: file                                Pv_M_I+4R_U_N_27d_1.txt  Rb_M_I_Fc-L_N_16d_3.txt  \
file
Pv_M_I+4R_U_N_27d_1.txt                    0.000000                0.825634
Rb_M_I_Fc-L_N_16d_3.txt                    0.825634                0.000000
Pv_M_I+4R_U_N_29d_3.txt                    0.186360                0.999453
Pv_M_I_Fc_Hf_38d_2.txt                    0.285853                0.556588
Pv_M_I_U_N_15d_1.txt                      0.550161                1.329045
...
Pv_M_5xI_U_N_21d_2.txt                    0.105457                0.927568
Pv_M_I_U_N_18d_6.txt                      0.436877                0.639594
Pv_L_I+4xR_U_N_78d_1.txt                  0.263120                1.066627
Pv_M_I_U_N_32d_3.txt                      0.230314                1.048077
Pv_M_I+4R_U_N_31d_1.txt                   0.042060                0.854447

file                                Pv_M_I+4R_U_N_29d_3.txt  Pv_M_I_Fc_Hf_38d_2.txt  \
file
Pv_M_I+4R_U_N_27d_1.txt                    0.186360                0.285853
Rb_M_I_Fc-L_N_16d_3.txt                    0.999453                0.556588
Pv_M_I+4R_U_N_29d_3.txt                    0.000000                0.470338
Pv_M_I_Fc_Hf_38d_2.txt                    0.470338                0.000000
Pv_M_I_U_N_15d_1.txt                      0.377332                0.828894
...
Pv_M_5xI_U_N_21d_2.txt                    0.111899                0.385329
Pv_M_I_U_N_18d_6.txt                      0.532882                0.409936
Pv_L_I+4xR_U_N_78d_1.txt                  0.199583                0.513373
Pv_M_I_U_N_32d_3.txt                      0.072756                0.515295
Pv_M_I+4R_U_N_31d_1.txt                   0.151238                0.321974

file                                Pv_M_I_U_N_15d_1.txt  Pv_M_I_Fc_Hf_22d_1.txt  \
file
Pv_M_I+4R_U_N_27d_1.txt                    0.550161                0.035963
Rb_M_I_Fc-L_N_16d_3.txt                    1.329045                0.848524
Pv_M_I+4R_U_N_29d_3.txt                    0.377332                0.177133
```

Pv_M_I_Fc_Hf_38d_2.txt	0.828894	0.302399
Pv_M_I_U_N_15d_1.txt	0.000000	0.545517
...
Pv_M_5xI_U_N_21d_2.txt	0.465073	0.087499
Pv_M_I_U_N_18d_6.txt	0.757071	0.471237
Pv_L_I+4xR_U_N_78d_1.txt	0.451924	0.232973
Pv_M_I_U_N_32d_3.txt	0.333874	0.217843
Pv_M_I+4R_U_N_31d_1.txt	0.510708	0.058807
file	Pv_M_I_Fc_Hf_22d_3.txt	Pv_M_I+R_U_N_25d_2.txt \
file		
Pv_M_I+4R_U_N_27d_1.txt	0.510857	0.683918
Rb_M_I_Fc-L_N_16d_3.txt	0.450957	0.205435
Pv_M_I+4R_U_N_29d_3.txt	0.650687	0.847215
Pv_M_I_Fc_Hf_38d_2.txt	0.362617	0.446525
Pv_M_I_U_N_15d_1.txt	0.923710	1.155070
...
Pv_M_5xI_U_N_21d_2.txt	0.600694	0.784287
Pv_M_I_U_N_18d_6.txt	0.192404	0.442198
Pv_L_I+4xR_U_N_78d_1.txt	0.764886	0.937136
Pv_M_I_U_N_32d_3.txt	0.693097	0.894882
Pv_M_I+4R_U_N_31d_1.txt	0.522942	0.706595
file	Pv_L_I+4xR_U_N_155d_1.txt	Rb_M_I_Fc-H_N_16d_3.txt \
file		
Pv_M_I+4R_U_N_27d_1.txt	0.224900	0.876111
Rb_M_I_Fc-L_N_16d_3.txt	1.044908	0.278068
Pv_M_I+4R_U_N_29d_3.txt	0.120490	1.026358
Pv_M_I_Fc_Hf_38d_2.txt	0.499113	0.659316
Pv_M_I_U_N_15d_1.txt	0.396883	1.304863
...
Pv_M_5xI_U_N_21d_2.txt	0.121736	0.972901
Pv_M_I_U_N_18d_6.txt	0.612275	0.553889
Pv_L_I+4xR_U_N_78d_1.txt	0.087742	1.132182
Pv_M_I_U_N_32d_3.txt	0.088713	1.073162
Pv_M_I+4R_U_N_31d_1.txt	0.204685	0.894013
file	... Pv_L_I+4xR_Fc_N_99d_3.txt \	
file	...	
Pv_M_I+4R_U_N_27d_1.txt	...	0.166148
Rb_M_I_Fc-L_N_16d_3.txt	...	0.892280
Pv_M_I+4R_U_N_29d_3.txt	...	0.252102
Pv_M_I_Fc_Hf_38d_2.txt	...	0.338369
Pv_M_I_U_N_15d_1.txt	...	0.606389
...
Pv_M_5xI_U_N_21d_2.txt	...	0.166587
Pv_M_I_U_N_18d_6.txt	...	0.593193

Pv_L_I+4xR_U_N_78d_1.txt	...	0.197739	
Pv_M_I_U_N_32d_3.txt	...	0.276801	
Pv_M_I+4R_U_N_31d_1.txt	...	0.190154	
file	Rb_M_I_Fc-H_N_20d_1.txt	Rb_M_I_Fc-M_N_31d_2.txt	\
file			
Pv_M_I+4R_U_N_27d_1.txt	0.855142	0.618249	
Rb_M_I_Fc-L_N_16d_3.txt	0.428068	1.351946	
Pv_M_I+4R_U_N_29d_3.txt	0.992467	0.451047	
Pv_M_I_Fc_Hf_38d_2.txt	0.669687	0.875011	
Pv_M_I_U_N_15d_1.txt	1.257518	0.329602	
...	
Pv_M_5xI_U_N_21d_2.txt	0.943964	0.561759	
Pv_M_I_U_N_18d_6.txt	0.562132	0.825626	
Pv_L_I+4xR_U_N_78d_1.txt	1.096581	0.571152	
Pv_M_I_U_N_32d_3.txt	1.036648	0.456839	
Pv_M_I+4R_U_N_31d_1.txt	0.870829	0.582019	
file	Pv_M_I+4R_U_N_15d_1.txt	Rb_M_I_Fc-L_N_31d_1.txt	\
file			
Pv_M_I+4R_U_N_27d_1.txt	0.079561	1.455208	
Rb_M_I_Fc-L_N_16d_3.txt	0.797104	0.894267	
Pv_M_I+4R_U_N_29d_3.txt	0.246994	1.568770	
Pv_M_I_Fc_Hf_38d_2.txt	0.244624	1.288085	
Pv_M_I_U_N_15d_1.txt	0.615036	1.767391	
...	
Pv_M_5xI_U_N_21d_2.txt	0.152573	1.542095	
Pv_M_I_U_N_18d_6.txt	0.475215	1.051327	
Pv_L_I+4xR_U_N_78d_1.txt	0.272600	1.711029	
Pv_M_I_U_N_32d_3.txt	0.285785	1.614433	
Pv_M_I+4R_U_N_31d_1.txt	0.120149	1.462166	
file	Pv_M_5xI_U_N_21d_2.txt	Pv_M_I_U_N_18d_6.txt	\
file			
Pv_M_I+4R_U_N_27d_1.txt	0.105457	0.436877	
Rb_M_I_Fc-L_N_16d_3.txt	0.927568	0.639594	
Pv_M_I+4R_U_N_29d_3.txt	0.111899	0.532882	
Pv_M_I_Fc_Hf_38d_2.txt	0.385329	0.409936	
Pv_M_I_U_N_15d_1.txt	0.465073	0.757071	
...	
Pv_M_5xI_U_N_21d_2.txt	0.000000	0.506170	
Pv_M_I_U_N_18d_6.txt	0.506170	0.000000	
Pv_L_I+4xR_U_N_78d_1.txt	0.171419	0.673984	
Pv_M_I_U_N_32d_3.txt	0.134457	0.569607	
Pv_M_I+4R_U_N_31d_1.txt	0.086053	0.434129	
file	Pv_L_I+4xR_U_N_78d_1.txt	Pv_M_I_U_N_32d_3.txt	\

file		
Pv_M_I+4R_U_N_27d_1.txt	0.263120	0.230314
Rb_M_I_Fc-L_N_16d_3.txt	1.066627	1.048077
Pv_M_I+4R_U_N_29d_3.txt	0.199583	0.072756
Pv_M_I_Fc_Hf_38d_2.txt	0.513373	0.515295
Pv_M_I_U_N_15d_1.txt	0.451924	0.333874
...
Pv_M_5xI_U_N_21d_2.txt	0.171419	0.134457
Pv_M_I_U_N_18d_6.txt	0.673984	0.569607
Pv_L_I+4xR_U_N_78d_1.txt	0.000000	0.170706
Pv_M_I_U_N_32d_3.txt	0.170706	0.000000
Pv_M_I+4R_U_N_31d_1.txt	0.254308	0.196891

file	Pv_M_I+4R_U_N_31d_1.txt
------	-------------------------

file	
Pv_M_I+4R_U_N_27d_1.txt	0.042060
Rb_M_I_Fc-L_N_16d_3.txt	0.854447
Pv_M_I+4R_U_N_29d_3.txt	0.151238
Pv_M_I_Fc_Hf_38d_2.txt	0.321974
Pv_M_I_U_N_15d_1.txt	0.510708
...	...
Pv_M_5xI_U_N_21d_2.txt	0.086053
Pv_M_I_U_N_18d_6.txt	0.434129
Pv_L_I+4xR_U_N_78d_1.txt	0.254308
Pv_M_I_U_N_32d_3.txt	0.196891
Pv_M_I+4R_U_N_31d_1.txt	0.000000

[270 rows x 270 columns]

```
[108]: # ps_symetric_dist.to_csv('path_score_distances_substrate.csv',
      ↪index_label='index', header=True)
# Drop the specified column
# to_drop = ['Rb_M_I_Fc-M_N_31d_3.txt'] # Column name to drop
# ps_symetric_dist_2 = ps_symetric_dist_2.drop(columns=to_drop)
# ps_symetric_dist_2 = ps_symetric_dist_2.drop(index=to_drop)
```

```
[109]: ps_symetric_dist_2.to_csv('path_score_distances_dropped.csv',
      ↪index_label='index', header=True)
```

```
[115]: # graph
color_mapping = {
    'Pp': 0,
    'Pv': 1,
    'Ag': 2,
    'Pi': 3,
    'Rb': 4,
    'Sc': 5,
```

```

    'none': 6
}

def get_leaf_color(filename):
    if filename[-4:] == '.txt':
        value = filename[0:2]
        return color_mapping[value]
    else:
        return color_mapping['none']

# Assuming 'df' contains attributes for each leaf node
# Pass the leaf attributes to the color calculation function
leaf_colors = [get_leaf_color(index_name) for index_name in ps_symetric_dist_2.
    ↪ index]

# Your distance matrix
# ps_symetric_dist = pd.read_csv('distances_ln.csv')
# ps_symetric_dist.set_index('index', inplace=True)
# ps_symetric_dist = distances.drop(distances.columns[[0]], axis=1)

tree = linkage(ps_symetric_dist_2, method='average', optimal_ordering=False) #
    ↪ Use 'average' method for UPGMA

# Plot dendrogram
plt.figure(figsize=(20, 20))
# dendrogram(tree, labels=distances.index, leaf_rotation=90)
dendrogram(tree, labels=ps_symetric_dist_2.index, leaf_rotation=0,
    ↪ orientation="left", color_threshold=240, above_threshold_color='grey')
plt.title('UPGMA Dendrogram Using Pathscore')
plt.xlabel('Data Points')
plt.ylabel('Distance')
# plt.show()

my_palette = plt.cm.get_cmap('RdYlGn', 7)

ax = plt.gca()
xlbls = ax.get_ymajorticklabels()
num=-1
for lbl in xlbls:
    val = get_leaf_color(lbl.get_text())
    lbl.set_color(my_palette(val))

# Show the graph
plt.show()

# Define labels for each color

```

```

color_labels = ['Pp', 'Pv', 'Ag', 'Pi', 'Rb', 'Sc']

# Create a dictionary to map colors to labels
color_map = {my_palette(i): label for i, label in enumerate(color_labels)}

# Display the colors and their corresponding labels
for color, label in color_map.items():
    print(f'{label}: {color}')

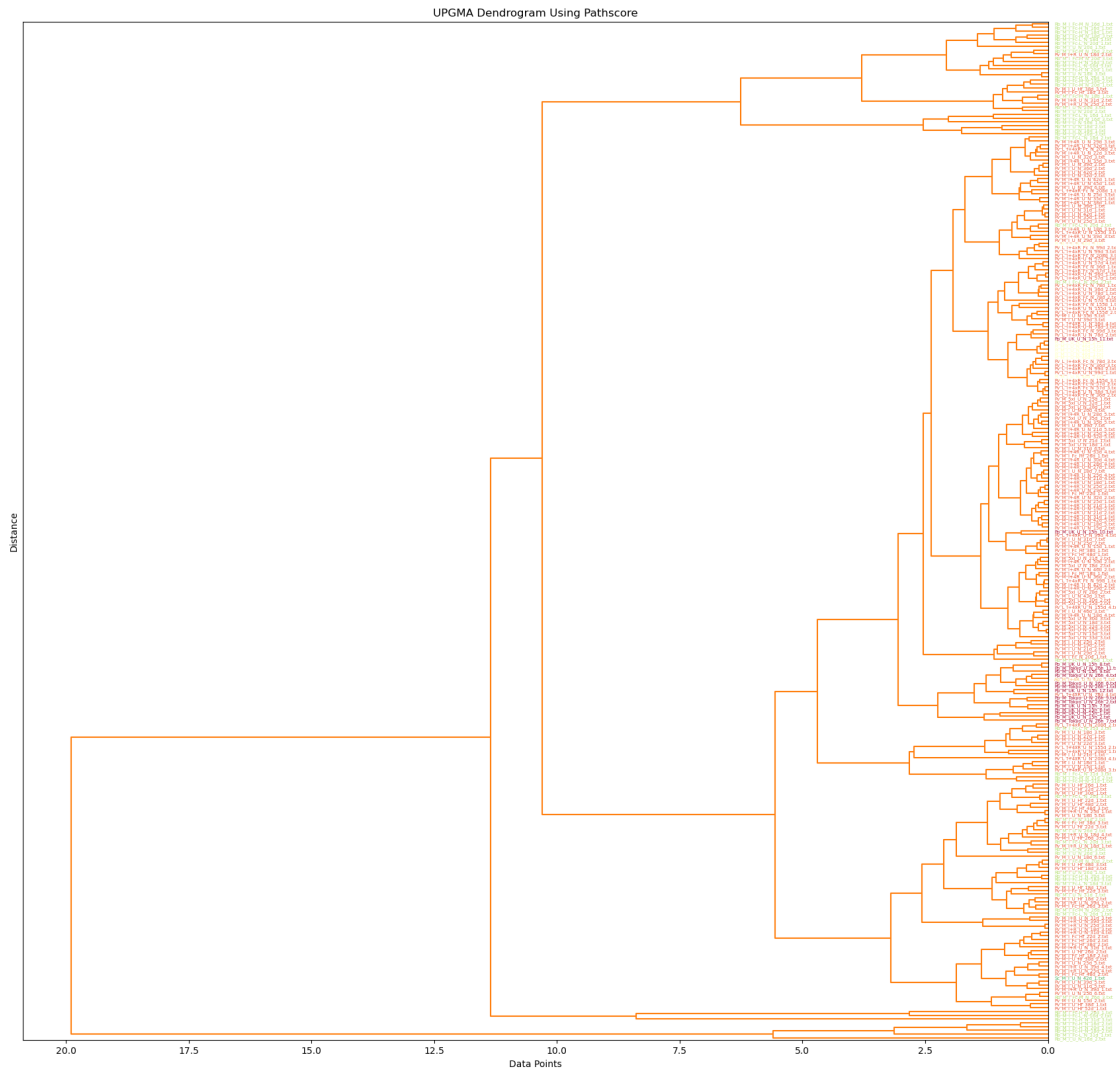
# Plot example colors
plt.figure(figsize=(6, 1))
for color, label in color_map.items():
    plt.plot([0, 1], [0, 0], color=color, label=label, linewidth=4)
plt.xlim(0, 1)
plt.ylim(-0.5, 0.5)
plt.yticks([])
plt.legend(loc='center', ncol=len(color_labels))
plt.title('Color Key')
plt.show()

```

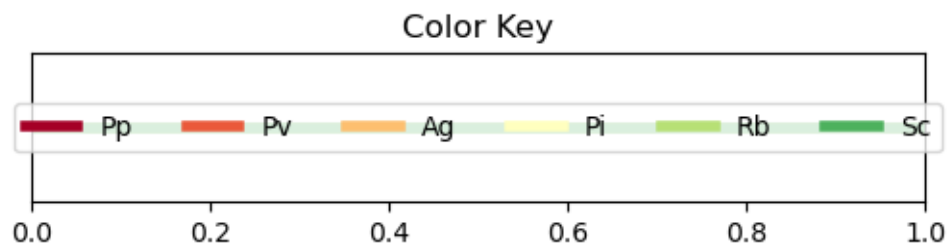
```

/tmp/ipykernel_375/1475335721.py:28: ClusterWarning: scipy.cluster: The
symmetric non-negative hollow observation matrix looks suspiciously like an
uncondensed distance matrix
    tree = linkage(ps_symetric_dist_2, method='average', optimal_ordering=False) #
Use 'average' method for UPGMA

```

Pp: (0.6470588235294118, 0.0, 0.14901960784313725, 1.0)
Pv: (0.918954248366013, 0.3477124183006535, 0.22614379084967318, 1.0)
Ag: (0.9934640522875817, 0.7477124183006535, 0.4352941176470587, 1.0)
Pi: (1.0, 1.0, 0.7490196078431373, 1.0)
Rb: (0.7176470588235295, 0.8797385620915033, 0.45882352941176474, 1.0)
Sc: (0.3006535947712421, 0.6928104575163401, 0.3633986928104576, 1.0)



```
[ ]: jupyter nbconvert --to pdf --TagRemovePreprocessor.enabled=True␣  
      ↪--TagRemovePreprocessor.remove_input_tags='{"graph"}' --output graphs.pdf␣  
      ↪Analysis.ipynb
```