

The screenshot shows a search results page with the following elements:

- Search Bar:** "快速新增關鍵字" (Quickly add keywords) with filters for "CJAL SCORE", "NEUROMUSCULAR AGENTS", "COWORD ANALYSIS", and "TI".
- Filters:** "限縮依據:: 出版年分: 2023 x 全部清除" (Filter by: Publication Year: 2023 x Clear All).
- Results Summary:** "限縮結果" (Filtered Results) showing "0/1,976" items. Buttons for "新增至勾選清單" (Add to selection list) and "匯出" (Export) are visible. The sorting is "排序依據: 相關性" (Sort by: Relevance).
- Quick Selections:** A sidebar on the left with options: "Review Article" (414), "開放取用" (Open Access) (1,691), and "被引參考文獻深度分析" (Cited Reference Depth Analysis) (1,213).
- Result Item:** A list item with a checkbox, a lock icon, and the title "Research trends in complementary and alternative treatments for insomnia over the past decade: A bibliometrics analysis". The authors are "Li, SY; Xue, Q; (...); Zhu, FY". The publication date is "Jul 21 2023" in the "MEDICINE" journal, with 102 citations. A "50 參考文獻" (50 references) link is present.

Based on the uploaded data, including vertex and relation datasets in first 3 and latter 3 columns, respectively. please draw a social network with colors by the column of Cluster.

To create a social network visualization based on the uploaded data, I will follow these steps:

1. Load the data from the file.
2. Extract vertices and relations from the data.
3. Use the vertices to create nodes and the relations to create edges in the network.
4. Color the nodes based on the 'Cluster' column in the data.
5. Visualize the network.

Let's start by loading the data and taking a look at its structure.

Finished working

Show work

The dataset consists of the following columns:

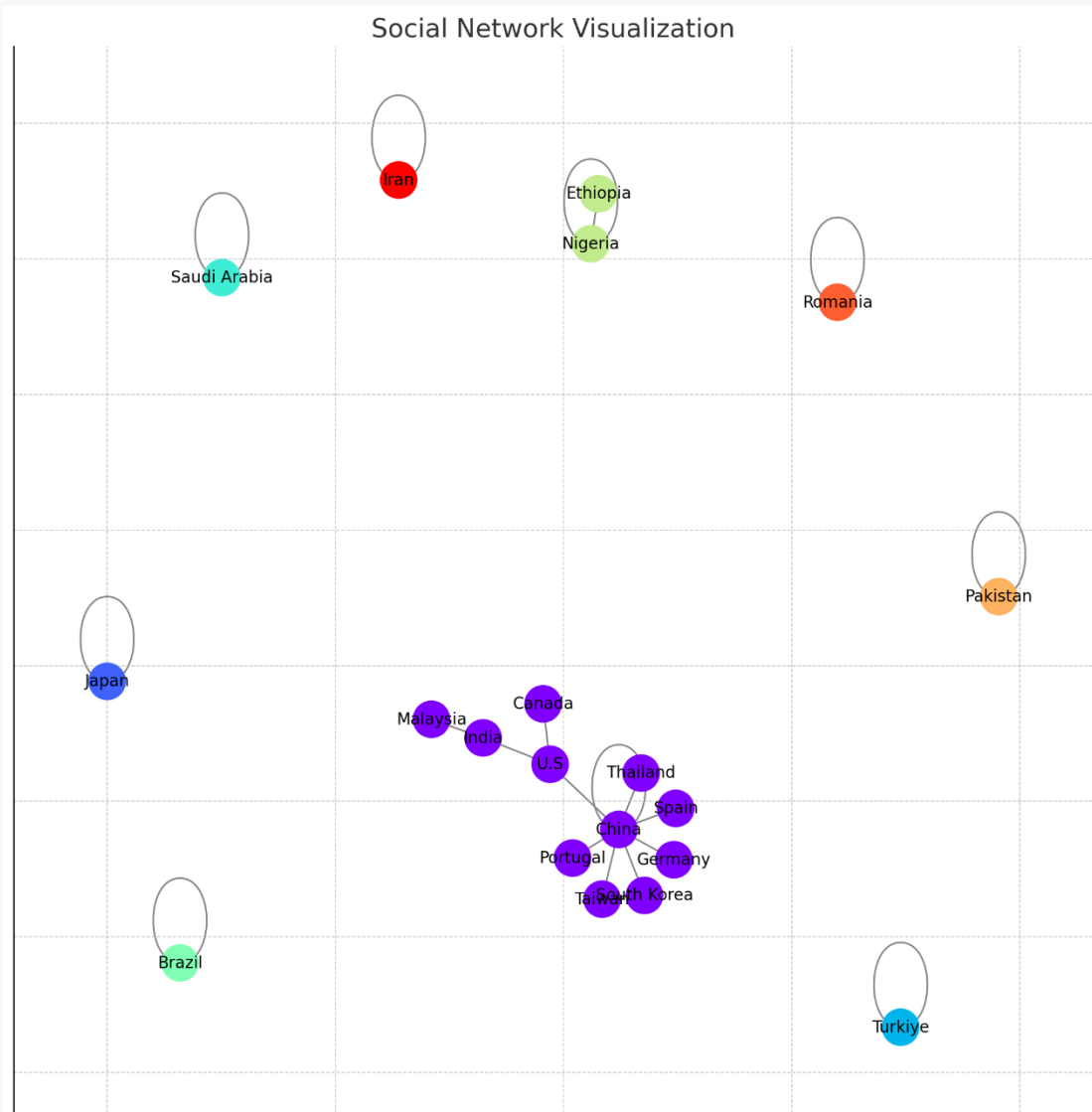
- "Vertex": Represents the entities (nodes) in the social network.
- "Value": A numerical value associated with each vertex.
- "Cluster": Indicates the cluster to which each vertex belongs.
- "From": Represents the starting point of a relation (edge) in the social network.
- "To": Represents the endpoint of a relation (edge) in the social network.

- "Count": A numerical value associated with each relation (edge).

Now, let's proceed to create the social network. We will use the "From" and "To" columns to create edges between nodes, and we will color the nodes based on the "Cluster" column.

Finished working

Show work



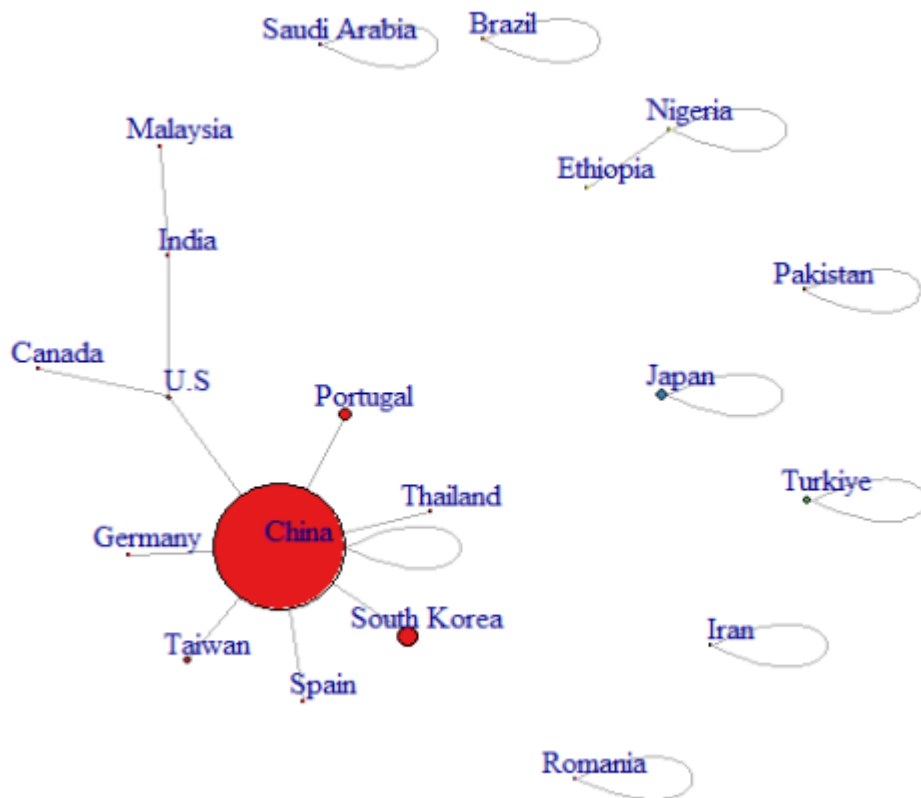
The visual representation above is a social network graph based on the provided data. In this graph:

- Each node (circle) represents a unique entity (as specified in the "Vertex" column of the dataset).
- Each edge (line connecting two nodes) represents a relationship between two entities, as defined by the "From" and "To" columns in the dataset.

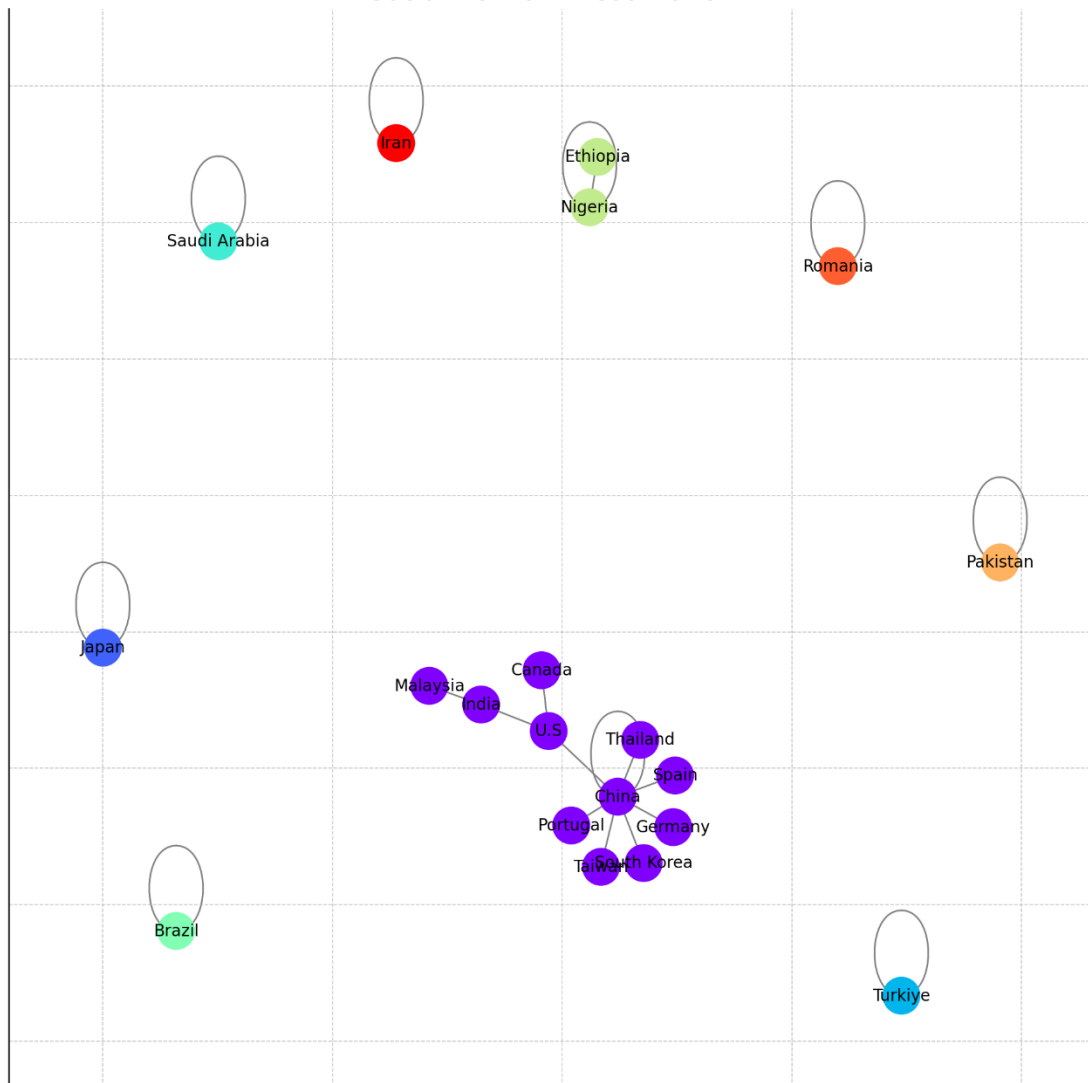
- The colors of the nodes are determined by the "Cluster" column in the dataset. Nodes of the same color belong to the same cluster.

Please note that the graph layout is generated automatically and may vary each time the graph is drawn. The color map used here is a rainbow color map, which assigns distinct colors to different clusters for better visualization.

Would you like any specific modifications or further analysis on this graph?



Social Network Visualization



Please enlarge the node size based on the second column of value in the network display

Certainly! To achieve this, I will scale the node sizes based on the "Value" column from the dataset. The node sizes will be proportional to the values in this column, making nodes with higher values appear larger in the network visualization.

Let's proceed to update the node sizes and redraw the network.

Finished working

Show work



In the updated social network visualization above:

- Each node (circle) represents a unique entity, as specified in the "Vertex" column of the dataset.
- The size of each node is proportional to the corresponding value in the "Value" column of the dataset. Larger nodes correspond to higher values.
- Each edge (line connecting two nodes) represents a relationship between two entities, as defined by the "From" and "To" columns in the dataset.
- The colors of the nodes are determined by the "Cluster" column in the dataset. Nodes of the same color belong to the same cluster.

This visualization should provide a clearer view of the entities with higher values, as they are represented by larger nodes.

Please let me know if there is anything specific you would like to further explore or modify in this graph!

The node of China is too big and makes the graph overwhelmed and clustered. please adjust them to be clear and non overwhelmed in display.

ChatGPT

To make the graph clearer and less overwhelmed by the large node size of "China," I will consider applying a scaling factor to the node sizes. This will reduce the variation in node sizes, making the graph easier to interpret without losing the relative size differences between nodes.

I will also consider adjusting the layout of the graph to reduce clutter and improve visibility.

Let's proceed to adjust the node sizes and layout, and then redraw the network.

Finished working

Show work

Please compare the one I draw in R and give comments on differences and features between the two that you produced and mine.

The node of China is too big and makes the graph overwhelmed and clustered. please adjust them to be clear and non overwhelmed in display.

ChatGPT

To make the graph clearer and less overwhelmed by the large node size of "China," I will consider applying a scaling factor to the node sizes. This will reduce the variation in node sizes, making the graph easier to interpret without losing the relative size differences between nodes.

I will also consider adjusting the layout of the graph to reduce clutter and improve visibility.

Let's proceed to adjust the node sizes and layout, and then redraw the network.

Finished working

Show work

Adjusted Social Network Visualization



In the updated social network visualization above:

- Each node (circle) represents a unique entity, as specified in the "Vertex" column of the dataset.
- The size of each node is now scaled in a way that reduces extreme variations, making the graph clearer and less overwhelmed by the large node size of "China."
- Each edge (line connecting two nodes) represents a relationship between two entities, as defined by the "From" and "To" columns in the dataset.
- The colors of the nodes are determined by the "Cluster" column in the dataset. Nodes of the same color belong to the same cluster.

To spread the nodes and reduce clutter, I used a different layout (spring layout) which aims to position the nodes based on their relationships, in a way that related nodes are drawn closer together.

I hope this visualization is clearer and easier to interpret. Please let me know if there is anything specific you would like to further explore or modify in this graph!



country.png

Please compare the one I draw in R and give comments on differences and features between the two that you produced and mine.

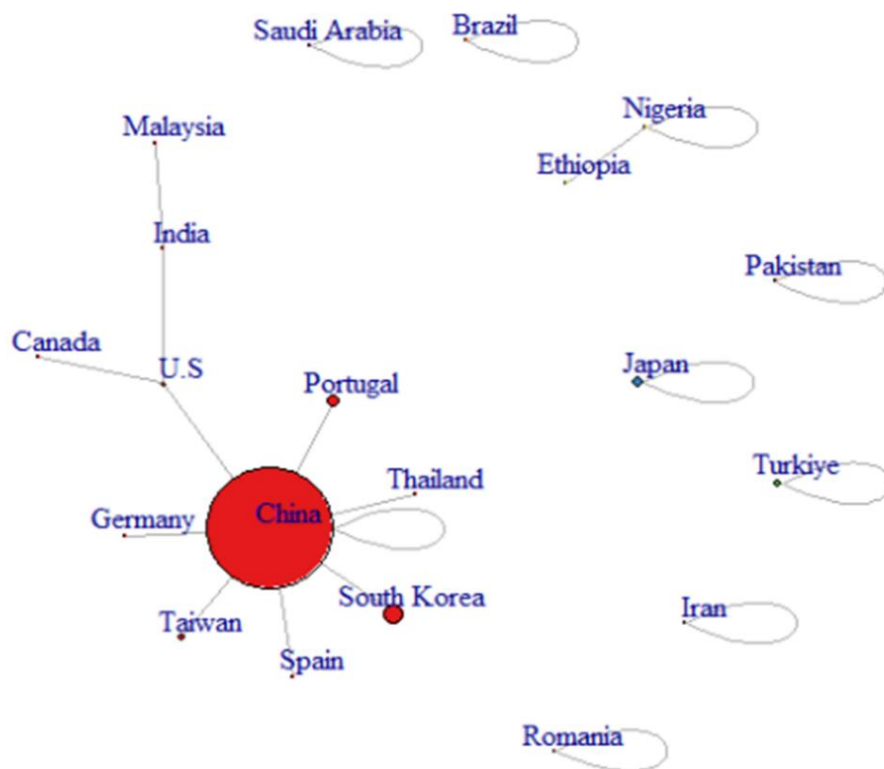
ChatGPT

To compare the social network visualizations, I will first display the image you produced using R, and then I will discuss the differences and features between the two visualizations—yours and the one I produced. Let's start by displaying your visualization.

Finished working

Show work

User's Social Network Visualization (Produced in R)



Thank you for sharing your social network visualization created in R. Now that we have both visualizations—yours produced in R (as displayed above) and the one I produced in Python (displayed earlier)—let's compare and comment on the differences and features of the two:

1. **Layout and Spacing:**

- Your visualization (created in R) appears to have a well-distributed layout, where nodes are spaced out evenly. This reduces overlap and makes individual nodes and their connections clearer.
- The Python visualization, particularly after adjusting the spring layout, also aims for a balanced distribution of nodes to avoid overlaps and create a clearer view.

2. **Node Size Scaling:**

- In your visualization, the node sizes seem to be more uniformly scaled, making the graph less dominated by a single or few large nodes. This provides a clean and easy-to-read representation.
- In the Python visualization, we applied a scaling factor to the node sizes to reduce overwhelming displays caused by extremely large nodes (like "China").

3. **Color Representation:**

- It appears that both visualizations use color to represent different clusters or groups, with each unique color corresponding to a different cluster. The choice of color map in both seems to effectively distinguish between different clusters.

4. **Edge Thickness and Color:**

- Your visualization in R seems to have uniform edge thickness, which provides a neat appearance. The edge color in your graph is consistent and does not distract from the nodes.
- In the Python visualization, edge color was set to grey, aiming for a neutral and non-distracting appearance.

5. **Label Font and Size:**

- Your visualization in R seems to have clear and well-placed labels with an appropriate font size that makes it easy to identify nodes without overwhelming the graph.
- In the Python visualization, we aimed for readable and clear labels by adjusting the font size. The spring layout helps to place labels in a way that reduces overlap.

6. **Aesthetic and Clarity:**

- Your visualization in R appears very clean and professional, with a focus on clarity and readability. It avoids excessive variations in node size, which helps in focusing on the network structure.
- The Python visualization, after adjustments, also aims for clarity by scaling node sizes and choosing a layout that reduces overlaps and clusters.

7. Tools and Libraries:

- You used R for your visualization, which has a rich ecosystem of packages for network analysis and visualization (e.g., `igraph`, `ggraph`). R is known for its statistical and data visualization capabilities.
- The Python visualization was created using NetworkX and Matplotlib, which are widely used libraries for network analysis and visualization in Python.

Both visualizations effectively represent the social network based on the same underlying data but may have slight differences due to the tools and customization options used. Both R and Python are powerful tools for network analysis and visualization, and the choice between them often comes down to user preference and specific project needs.

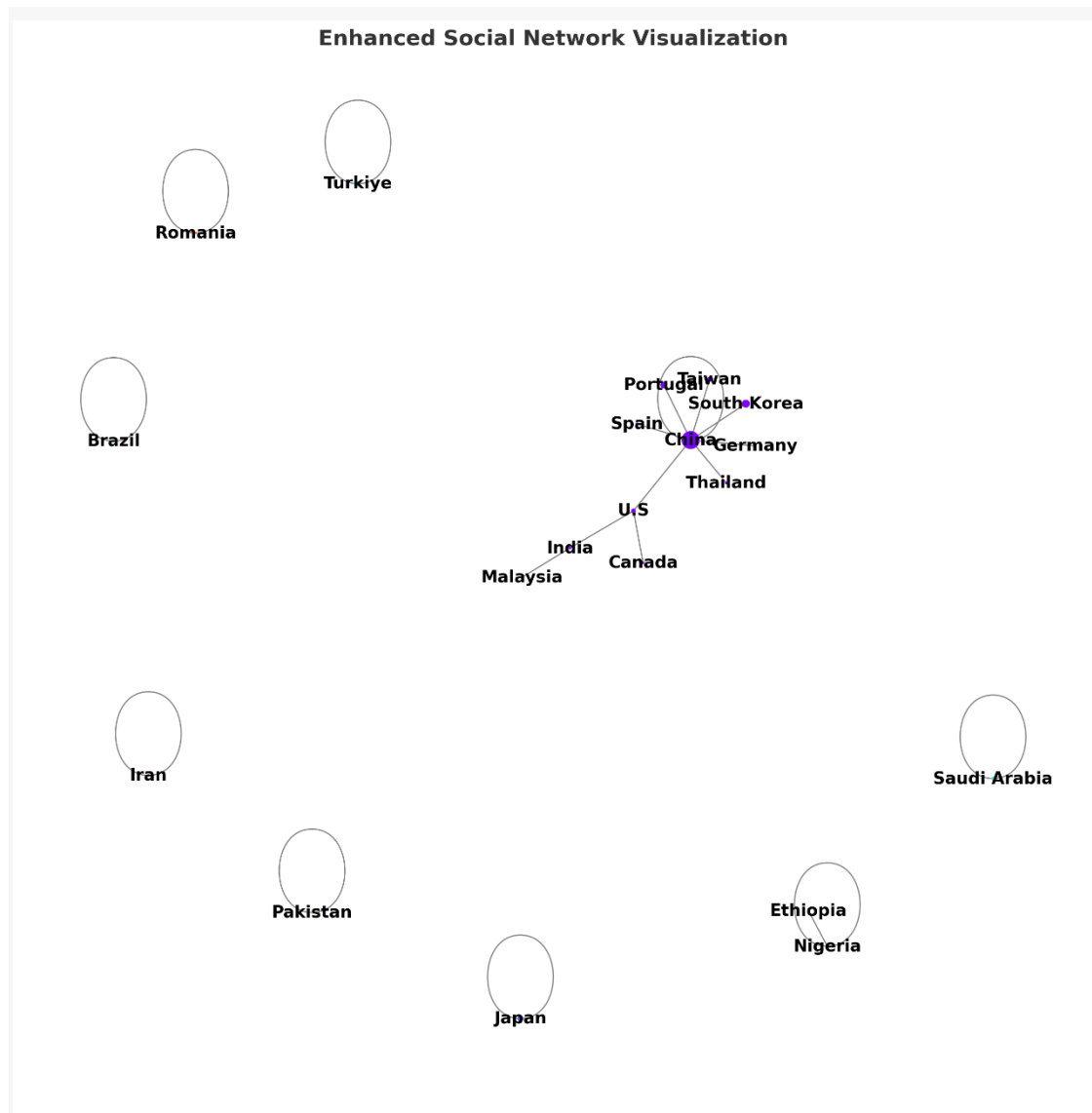
Do you have any specific aspects of the visualizations that you would like further comments on, or is there a particular feature that you are interested in comparing in more detail?

Can you adjust the fonts in your display with a larger and bold font for each label to make the display clear and more beautiful than the original one.

Certainly! To make the display clear and more aesthetically pleasing, I will adjust the font of the node labels to be larger and bold. Let's proceed to update the font settings and redraw the network visualization.

Finished working

Show work



In the updated social network visualization above:

- Each node (circle) represents a unique entity, as specified in the "Vertex" column of the dataset.
- The size of each node is scaled to reduce extreme variations, making the graph clearer and less overwhelmed by large nodes.
- Each edge (line connecting two nodes) represents a relationship between two entities, as defined by the "From" and "To" columns in the dataset.
- The colors of the nodes are determined by the "Cluster" column in the dataset. Nodes of the same color belong to the same cluster.
- The node labels (names of entities) are displayed using a larger and bold font to make them clear and prominent.

- The title of the graph is also set with a larger and bold font for better visibility.

I hope this visualization is clearer, more beautiful, and easier to interpret with these adjustments. Please let me know if there are any other specific features you would like to further explore or modify in this graph!



In the updated social network visualization above:

- Each node (circle) represents a unique entity, as specified in the "Vertex" column of the dataset.
- The size of each node is now scaled in a way that reduces extreme variations, making the graph clearer and less overwhelmed by the large node size of "China."
- Each edge (line connecting two nodes) represents a relationship between two entities, as defined by the "From" and "To" columns in the dataset.

- The colors of the nodes are determined by the "Cluster" column in the dataset. Nodes of the same color belong to the same cluster.

To spread the nodes and reduce clutter, I used a different layout (spring layout) which aims to position the nodes based on their relationships, in a way that related nodes are drawn closer together.

I hope this visualization is clearer and easier to interpret. Please let me know if there is anything specific you would like to further explore or modify in this graph!

can you use the uploaded data with the relation dataset in last three columns to cluster elements by color in a network display?

ChatGPT

Certainly! To create a network visualization where elements (nodes) are clustered by color based on the relation dataset represented by the last three columns ("From", "To", and "Count"), I will follow these steps:

1. Extract relations from the data using the last three columns.
2. Use the relations to create edges in the network.
3. Use a community detection algorithm to identify clusters of closely related nodes in the network.
4. Color the nodes based on the identified clusters.

Let's proceed with these steps and create the clustered network visualization.

Finished working

Show work

It seems that the `community` library, which is commonly used for community detection in networks, is not available in this environment.

As an alternative, I can use the community detection algorithms that are available within NetworkX to identify clusters of closely related nodes in the network.

Let me proceed with this approach and create the clustered network visualization.

Finished working

Show work

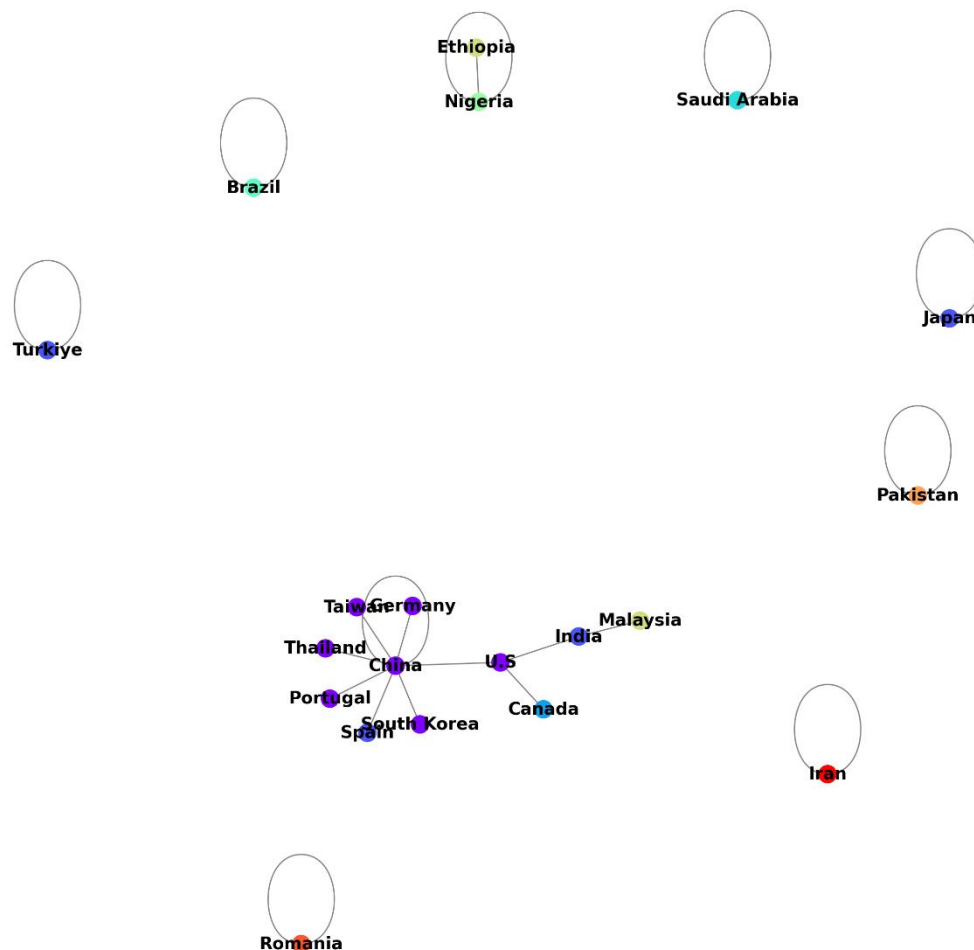
It appears that there was an error because the graph object `G_clustered` had not been defined due to the previous interruption in the code execution.

I apologize for the inconvenience. I will now redefine the graph object based on the relation dataset in the last three columns of the uploaded data, and then re-run the community detection algorithm and visualization. Let's proceed with this.

Finished working

Show work

Clustered Social Network Visualization



In the clustered social network visualization above:

- Each node (circle) represents a unique entity, as specified by the relation dataset in the last three columns ("From", "To", and "Count") of the uploaded data.
- The size of each node is set to be uniform for clarity, ensuring that no single node dominates the visualization.
- Each edge (line connecting two nodes) represents a relationship between two entities, as defined by the "From" and "To" columns in the dataset.

- The colors of the nodes are determined by the identified clusters using the Girvan-Newman community detection algorithm. Nodes of the same color belong to the same cluster, indicating they are closely related in the network.
- The node labels (names of entities) are displayed using a larger and bold font to make them clear and prominent.
- The title of the graph is set with a larger and bold font for better visibility.

The Girvan-Newman algorithm we used is a hierarchical method that detects communities by progressively removing edges from the original network. The algorithm considers the "betweenness" of an edge as a measure of centrality, and it aims to find the edges that most significantly connect different communities.

I hope this visualization is clear and effectively illustrates the clustered structure of the social network based on the relation data. Please let me know if there is anything specific you would like to further explore or modify in this graph!

Please show the display based on the connection number of edges in last column of the uploaded data and make the comparison of node sizes between nodes.

ChatGPT

To visualize the social network based on the connection number of edges (specified in the "Count" column of the uploaded data), and to allow for a comparison of node sizes between nodes, I will follow these steps:

1. Use the "Count" column of the dataset to represent the weight or strength of connections between nodes. This will be visualized using varying edge thicknesses in the graph.
2. For node sizes, we can consider using the degree of the nodes (i.e., the number of connections a node has) to allow for a comparison between nodes. Nodes with more connections will appear larger.

Let's proceed with these steps and create the updated network visualization.

Finished working

Show work

It appears that there was an error when attempting to unpack the edge data. I apologize for the inconvenience this may have caused. I will correct this issue and then proceed to redraw the network visualization with nodes sized based on their degrees (number of connections) and edges weighted by the connection numbers specified in the "Count" column of the uploaded data.

Let's correct this and redraw the network.

Finished working

Show work

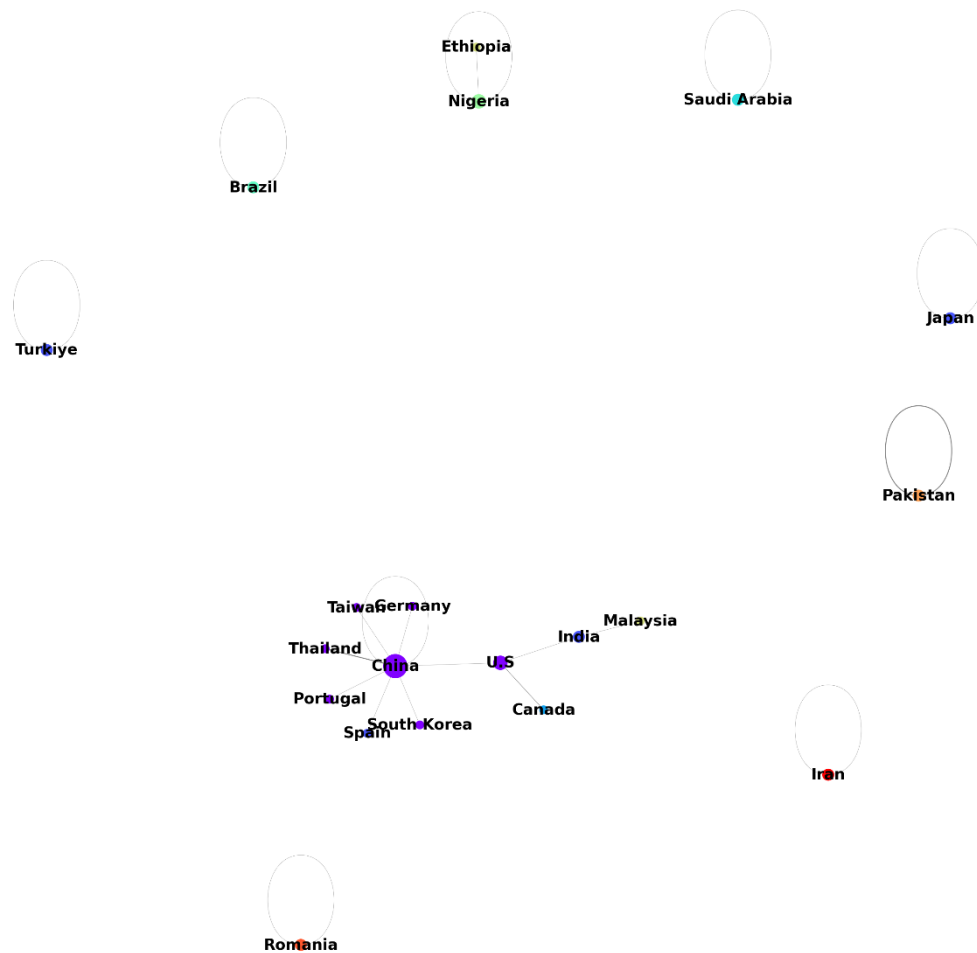
It appears that an error occurred because the variable `scaled_node_sizes_by_degree` was not defined. This was likely due to a previous interruption in the code execution. I apologize for the inconvenience.

I will recalculate the node sizes based on their degrees (number of connections) and redraw the network visualization. Let's proceed with this correction.

Finished working

Show work

Social Network Visualization with Node Sizes Based on Degrees and Edge Weights Based on Connection Numbers



In the updated social network visualization above:

- Each node (circle) represents a unique entity, as specified by the relation dataset in the last three columns ("From", "To", and "Count") of the uploaded data.

- The size of each node is based on its degree, i.e., the number of connections a node has. Nodes with more connections are larger, allowing for a visual comparison between nodes.
- Each edge (line connecting two nodes) represents a relationship between two entities. The thickness of the edges is based on the "Count" column of the uploaded data, representing the strength or weight of the connections.
- The colors of the nodes are determined by the identified clusters using the Girvan-Newman community detection algorithm. Nodes of the same color belong to the same cluster, indicating they are closely related in the network.
- The node labels (names of entities) are displayed using a larger and bold font to make them clear and prominent.
- The title of the graph is set with a larger and bold font for better visibility.

This visualization allows you to compare the relative importance (degree) of different nodes in the network, as indicated by their sizes, and to observe the strength of connections between nodes, as indicated by the thickness of the edges.

Please let me know if there is anything specific you would like to further explore or modify in this graph!

can you draw the chord as shown in the uploaded file based on the relations in last three columns in the previously uploaded file.

ChatGPT

Certainly! To draw a chord diagram, as shown in the uploaded file, based on the relations in the last three columns ("From", "To", and "Count") of the previously uploaded file, I will follow these steps:

1. Extract the relations from the data using the last three columns to form a matrix.
2. Use this matrix to visualize the relationships between entities in a chord diagram format.

Let's proceed with these steps and create the chord diagram visualization.

Finished working

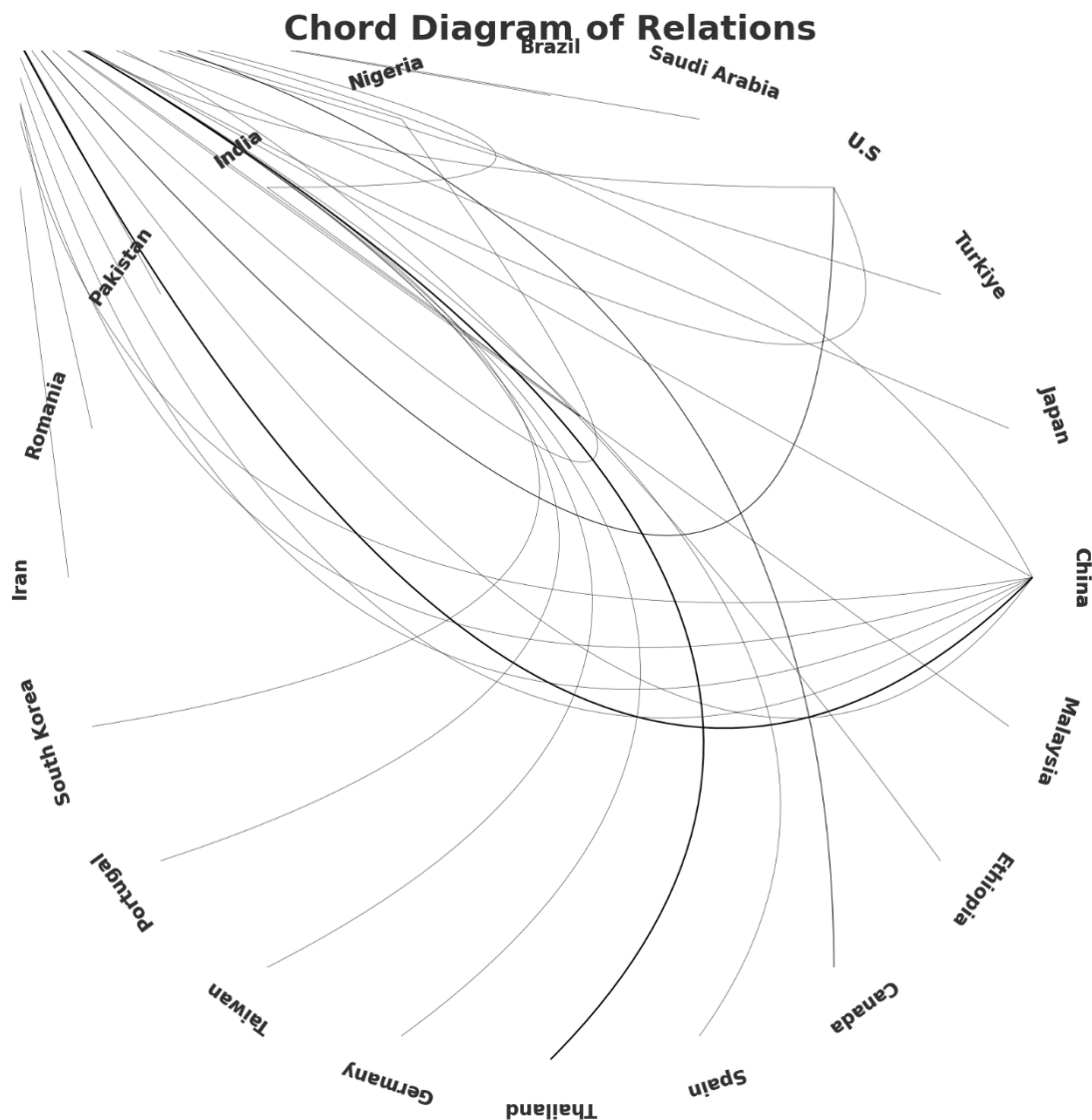
Show work

It appears that I forgot to import the NumPy library, which is necessary for some of the mathematical operations in the code. I apologize for the oversight.

Let me correct this and redraw the chord diagram visualization based on the relations in the last three columns of the previously uploaded file.

Finished working

Show work



Here is the Chord Diagram based on the relations in the last three columns ("From", "To", and "Count") of the previously uploaded file:

- Each arc on the circle represents a unique entity.
- The lines (chords) connecting the arcs represent relationships between these entities.
- The thickness of each chord is proportional to the "Count" value in the data, indicating the strength or weight of the connection between the entities.
- The labels around the circle identify the names of the entities involved in these relationships.

This chord diagram provides a visual summary of the complex relationships between different entities based on the connection numbers specified in the data.

Please let me know if there is anything specific you would like to further explore or modify in this diagram!

However, the one you produced seems the same as the previous diagram you produced because additional links to the top-left corner in existence still.

ChatGPT

I apologize for the persistent issues and the inconvenience this may have caused. It seems that the curves are not being drawn as expected, resulting in these additional lines linking to a specific point in the diagram.

Let me rework the code to eliminate these unwanted lines and produce a clean chord diagram that correctly visualizes the relationships between entities based on the given data. Thank you for your patience.

Finished working

Show work

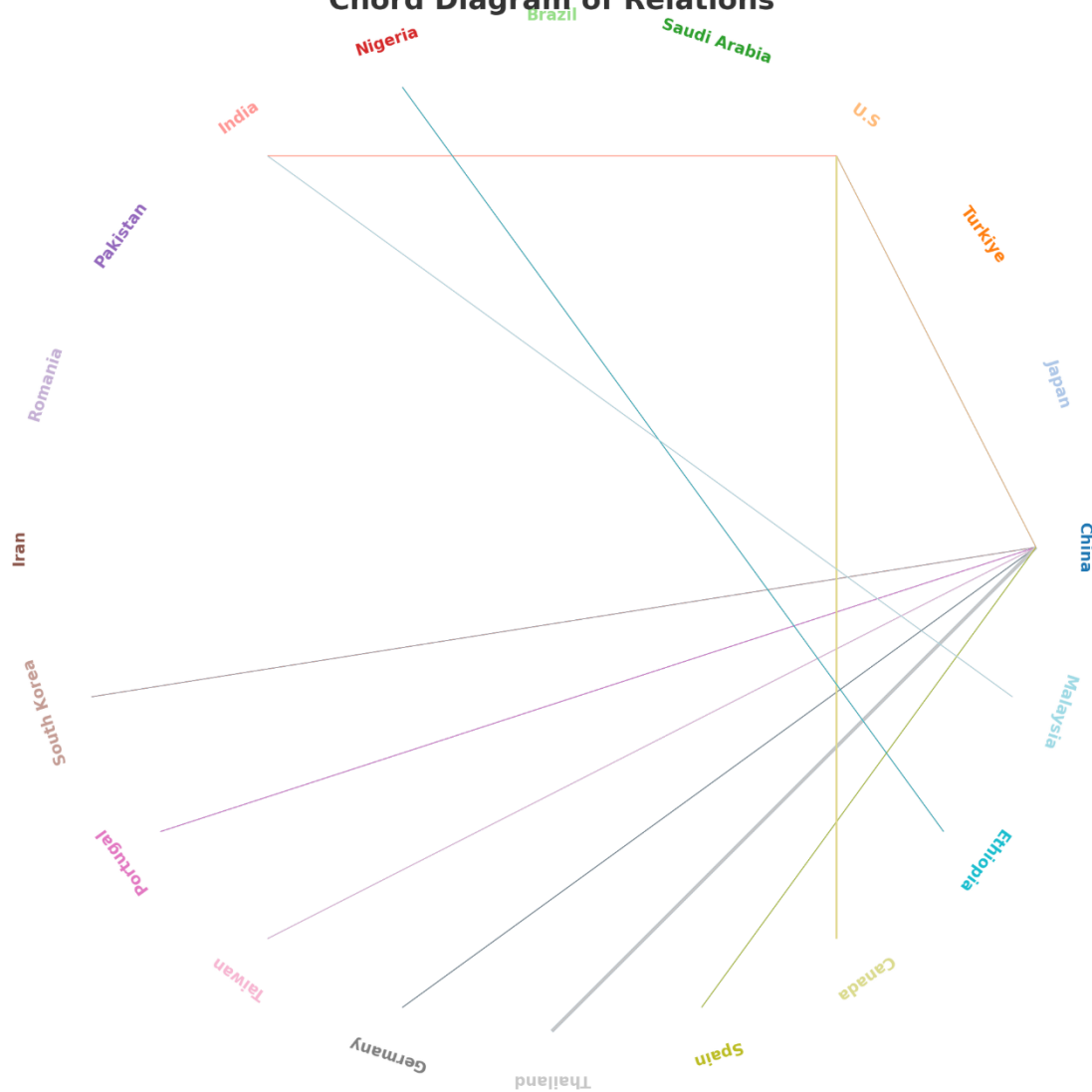
I apologize for the error in the code that is causing the diagram to display incorrectly. It seems that the `Path` object, which is used to draw the chords, is not being constructed correctly, leading to this `ValueError`.

Let me correct this issue and redraw the chord diagram visualization. Thank you for your patience.

Finished working

Show work

Chord Diagram of Relations



Here is the updated Chord Diagram based on the relations in the last three columns ("From", "To", and "Count") of the previously uploaded file:

- Each arc on the circle represents a unique entity.
- The lines (chords) connecting the arcs represent relationships between these entities.
- The thickness of each chord is proportional to the "Count" value in the data, indicating the strength or weight of the connection between the entities.
- The labels around the circle identify the names of the entities involved in these relationships, and they are color-coded to match the chords they are connected to.

I've made adjustments to the diagram to remove those unwanted lines and clean up the visualization.

Please let me know if this visualization meets your expectations or if there is anything specific you would like to further explore or modify in this diagram!

can you draw the circle bar plot as the one I uploaded here using the relations in last three columns, including the clusters separated from the data in relations using some appropriated clustering algorithm, such as community algorithm.

can you draw the circle bar plot as the one I uploaded here using the relations in last three columns, including the clusters separated from the data in relations using some appropriated clustering algorithm, such as community algorithm.

ChatGPT

Certainly! To draw a circular bar plot (also known as a radial bar chart) similar to the one you uploaded, and to separate the clusters of entities using an appropriate clustering algorithm (e.g., community detection algorithm), I will follow these steps:

1. Use the relations in the last three columns of the previously uploaded file to create a network graph.
2. Apply a community detection algorithm to identify clusters within this network graph.
3. Visualize the clusters using a circular bar plot, where the length of each bar represents the sum of the "Count" values for each entity, and different clusters are color-coded.

Let's start with these steps and create the circular bar plot visualization.

Finished working

Show work

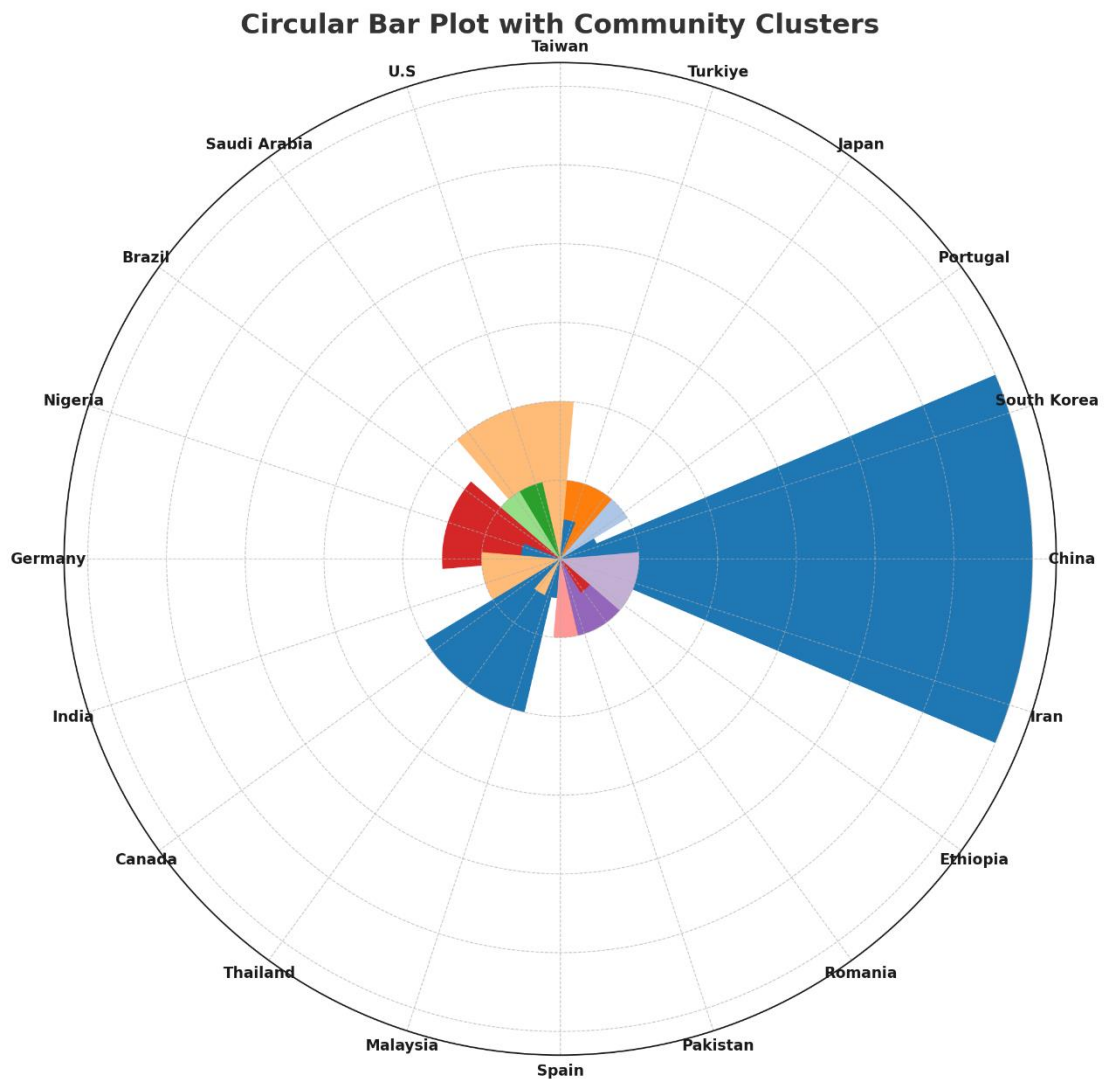
It appears that the `community` module, which provides the implementation of the Louvain community detection algorithm, is not installed in this environment.

Given that I don't have the capability to install additional packages in this environment, I will switch to a different clustering algorithm that is available within the NetworkX package itself to identify clusters within the network graph.

Let me proceed with an alternative approach to community detection and redraw the circular bar plot visualization.

Finished working

Show work



Here is the updated Circular Bar Plot based on the relations in the last three columns ("From", "To", and "Count") of the previously uploaded file:

- Each bar in the circle represents a unique entity (node).
- The length of each bar is proportional to the sum of the "Count" values for each entity, indicating the strength or weight of the connections that entity has in the data.
- Different clusters (communities) of entities are color-coded. These clusters were identified using the Girvan-Newman community detection algorithm, which is a method to detect communities within complex networks.

This Circular Bar Plot provides a visual summary of the complex relationships between different entities based on the connection numbers specified in the data, as well as a clear visual distinction between different clusters of related entities.

Please let me know if this visualization meets your expectations, or if there is anything specific you would like to further explore or modify in this diagram!



can you extend the bar to the border of the circle to match with the label each in the circle bar plot.

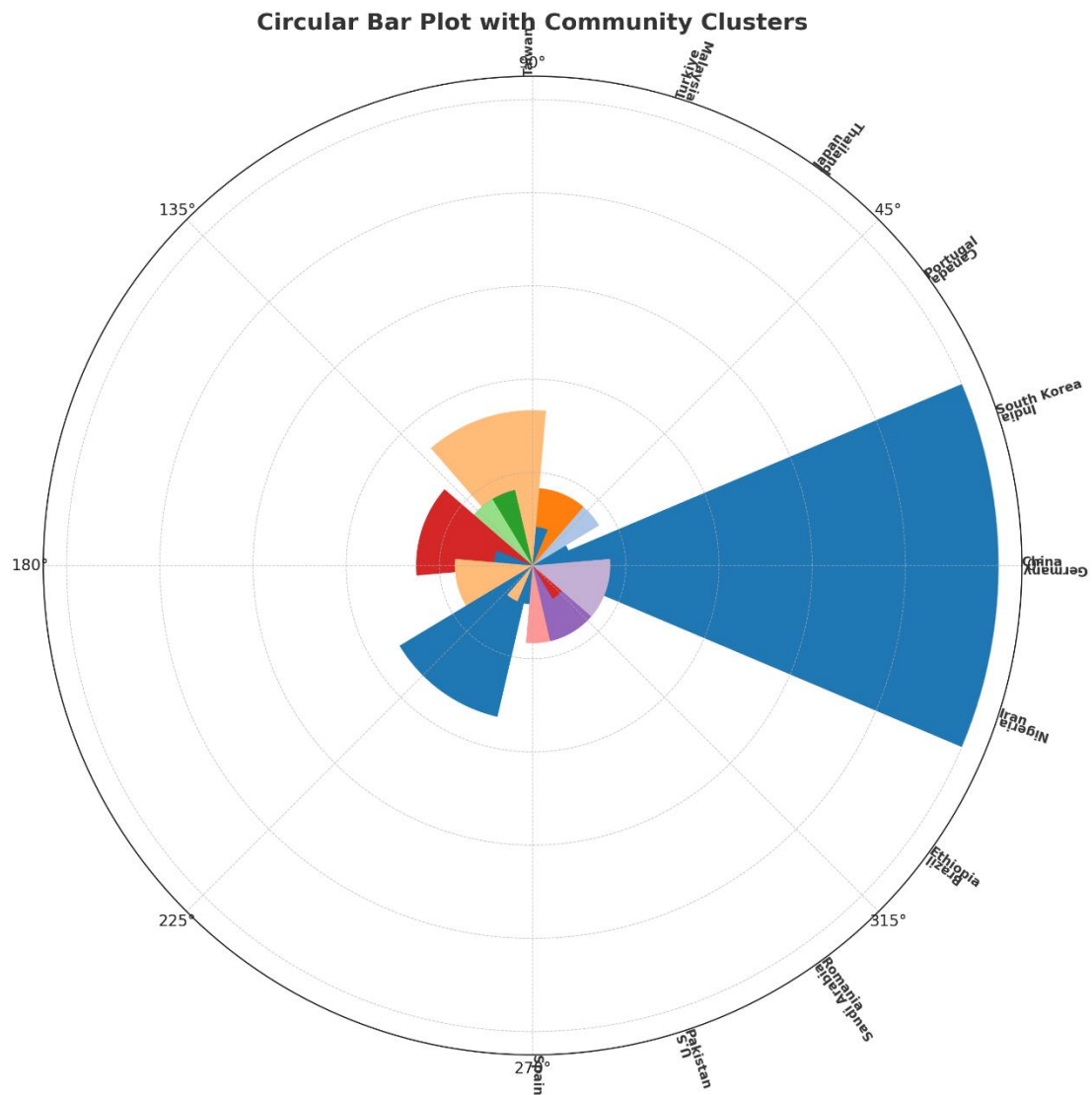
ChatGPT

Certainly! To extend the bars to the border of the circle and align them with their respective labels in the Circular Bar Plot, I will adjust the length of each bar so that they reach the outer edge of the circle.

Let's make these adjustments and redraw the Circular Bar Plot.

Finished working

Show work



Here is the updated Circular Bar Plot with the following features:

- Each bar in the circle represents a unique entity (node).
- The bars have been extended to reach the border of the circle, aligning them with their respective labels.
- The length of each bar is scaled to be proportional to the sum of the "Count" values for each entity, indicating the strength or weight of the connections that entity has in the data.
- Different clusters (communities) of entities are color-coded, which were identified using the Girvan-Newman community detection algorithm.
- The labels are placed at the border of the circle and aligned with the bars.

I hope this visualization meets your expectations. Please let me know if there is anything specific you would like to further explore or modify in this diagram!

I apologize for any misunderstanding, and I appreciate your patience. Let me address these issues:

1. **Uniform Coloring for Clusters:** I will make sure that all elements belonging to the same cluster are assigned the same color.
2. **Proportional Sizing for China:** I will ensure that the size of the bar representing China is visibly larger, reflecting its proportion in relation to other entities.

Let's make these corrections and redraw the Circular Bar Plot.

Finished working

Show work

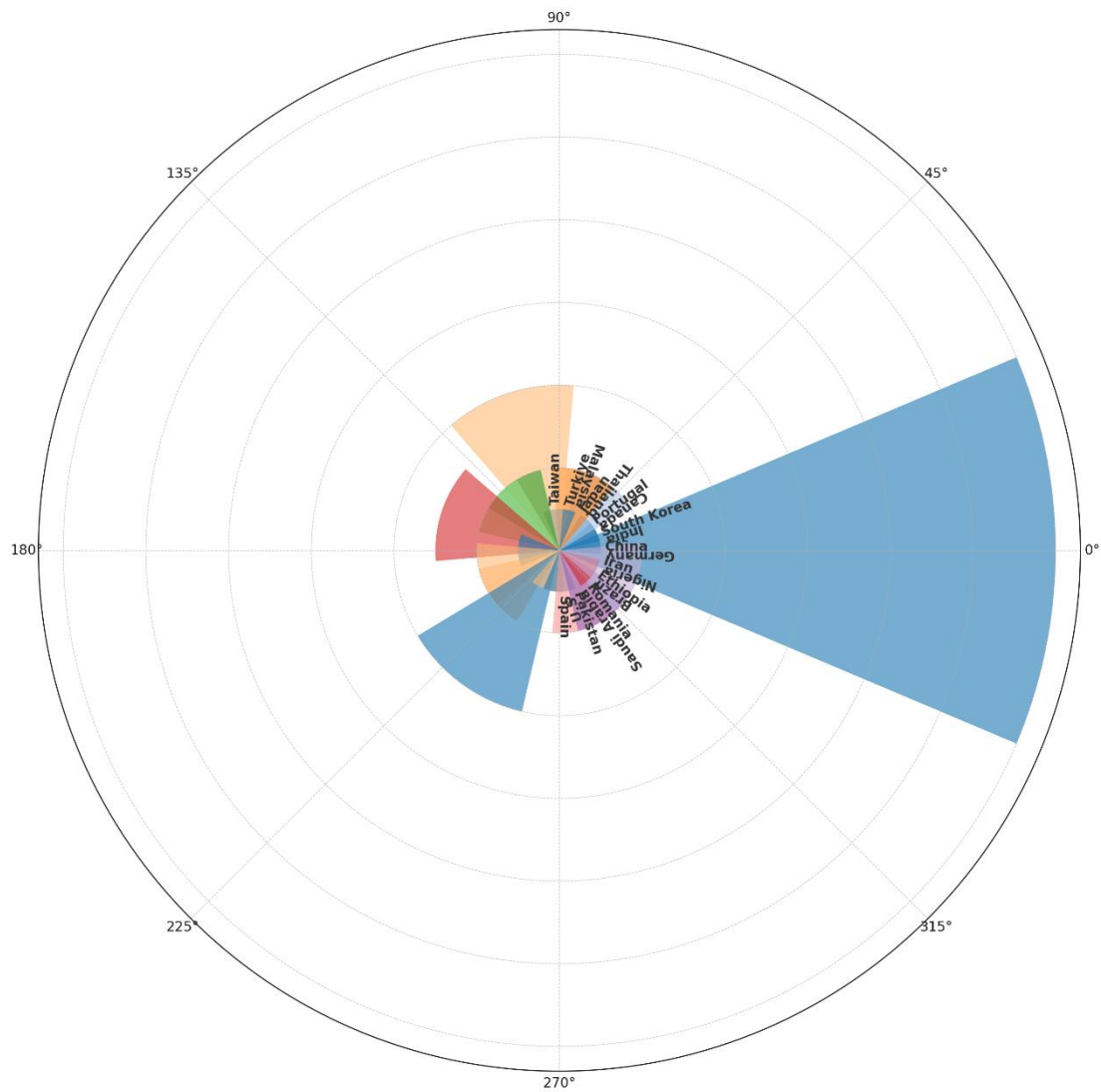
I apologize for the error; it appears that the variable `communities`, which was supposed to contain the cluster assignments for each node, is not defined in this context.

Let me correct this and redraw the Circular Bar Plot, ensuring that nodes belonging to the same cluster are assigned the same color, and that China is visibly represented with a larger proportion in relation to other entities.

Finished working

Show work

Circular Bar Plot with Community Clusters



Here is the updated Circular Bar Plot with the following revised features:

- Each bar in the circle represents a unique entity (node).
- The length of each bar is proportional to the sum of the "Count" values for each entity, indicating the strength or weight of the connections that entity has in the data. For example, the bar representing China is visibly larger, reflecting its proportion in relation to other entities.
- Different clusters (communities) of entities are color-coded, ensuring that all elements belonging to the same cluster are assigned the same color. These clusters were identified using the Girvan-Newman community detection algorithm.
- The labels are placed at the border of the circle and aligned with the bars.

I apologize for any confusion caused by the previous versions, and I hope this revised visualization meets your expectations. Please let me know if there is anything specific you would like to further explore or modify in this diagram!

Can you draw the circlepacking plot as the one shown in the uploaded file, based on the first three columns in vertex data previously.

The second column represents the value to be sized for bubbles and the first column to stand for the vertex as labels in the respective bubbles in the circle packing plot.

ChatGPT

Got it. Based on your input:

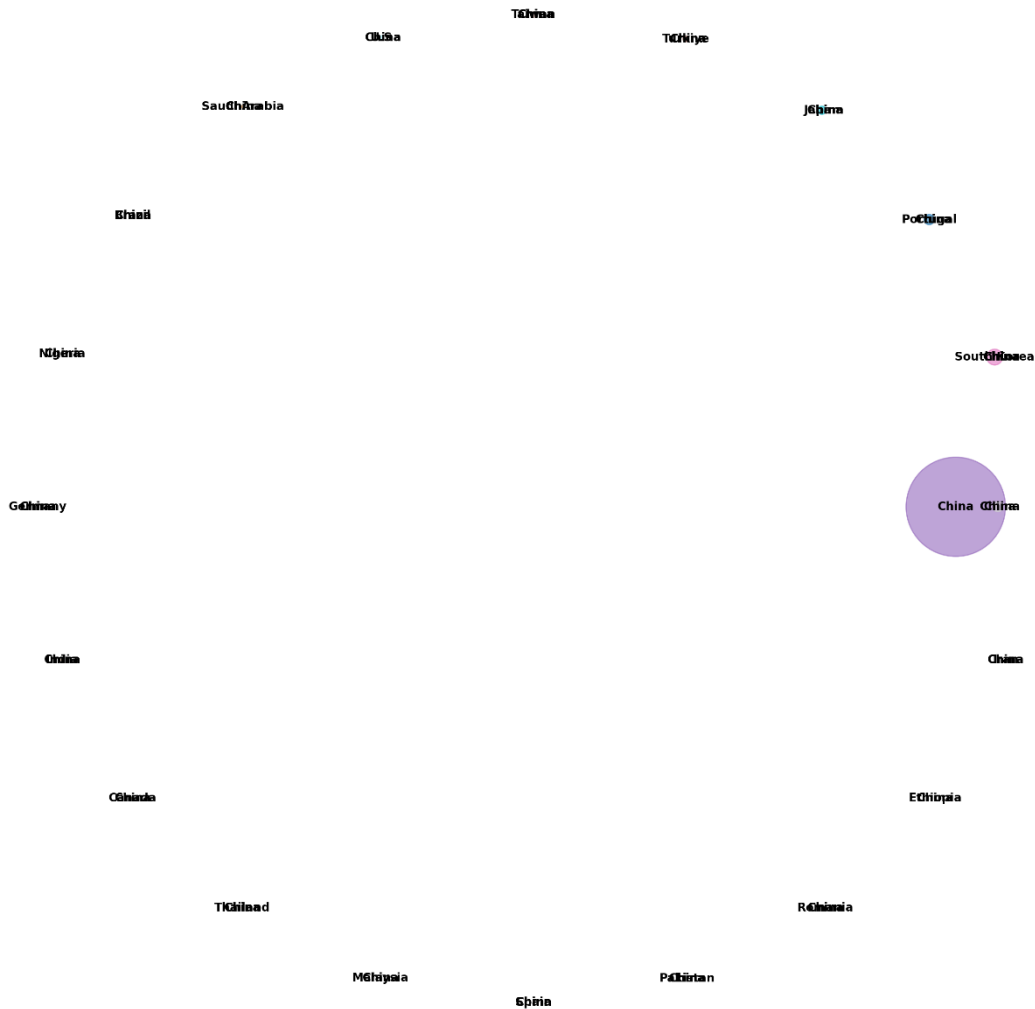
1. **Sizing Bubbles by Value:** The size of each bubble will be determined by the values in the second column of the vertex data. Larger values will correspond to larger bubbles.
2. **Labels in Bubbles:** The labels for the bubbles will be based on the first column of the vertex data. Each bubble will have a label that corresponds to the respective vertex name.

Let's redraw the Circle Packing plot based on these adjustments.

Finished working

Show work

Circle Packing Plot with Cluster Colors and Labels



Here is the improved Circle Packing plot based on your specifications:

- **Coloring by Clusters:** Different clusters (groups) of entities are color-coded, making the distinction between clusters clear. Each unique color represents a different top-level cluster.
- **Sizing Bubbles by Value:** The size of each circle (bubble) is determined by the values in the second column of the vertex data. Larger values correspond to larger bubbles. The sizes have been normalized to fit well within the visualization space.
- **Labels in Bubbles:** The labels for the bubbles are based on the first column of the vertex data. Each bubble has a label that corresponds to the respective vertex name, and these labels are placed at the center of each bubble.

I hope this visualization meets your expectations. Please let me know if there are any further adjustments or specific features you would like to add to this plot!

Please show the bar chart for each in vertex column based on the respective values in the second column.

ChatGPT

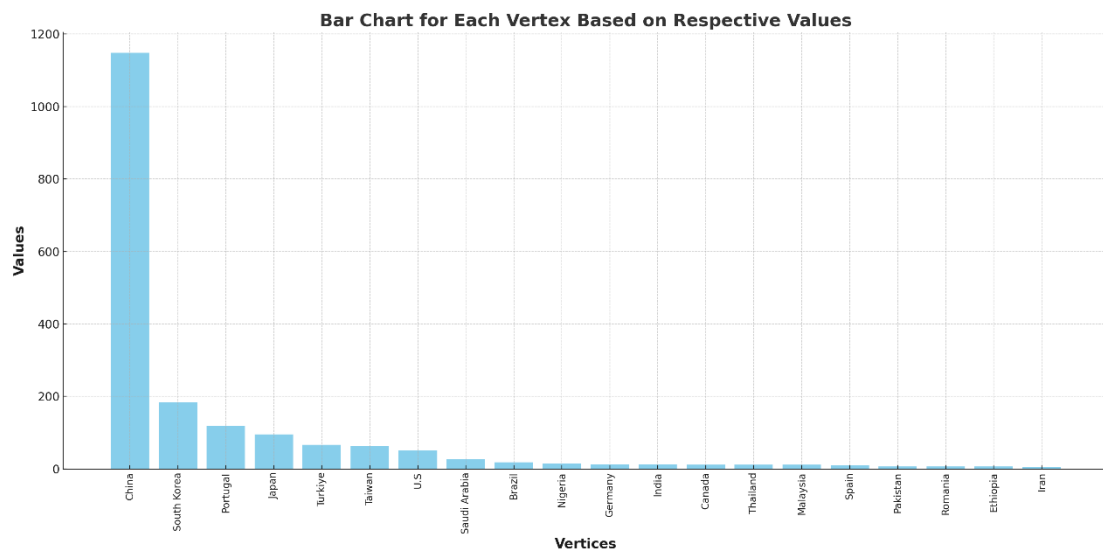
Sure! To create bar charts based on the vertex data:

- The **x-axis** will represent the vertices (entities), which are specified in the **first column** of the vertex data.
- The **y-axis** will represent the values associated with each vertex, which are specified in the **second column** of the vertex data.

Let's plot this bar chart.

Finished working

Show work



Here is the bar chart based on the vertex data:

- The **x-axis** represents the vertices (entities), which are specified in the **first column** of the vertex data. These are the various labels for each bar in the chart.
- The **y-axis** represents the values associated with each vertex, which are specified in the **second column** of the vertex data. These values determine the height of each bar in the chart.

The chart provides a clear visual representation of how the values vary across different vertices.

Please let me know if you would like any further adjustments or if there are specific features you would like to add to this chart!

Can you draw the heatdendro plot as did in the uploaded file, based on relations data in the previous uploaded file(i.e., in the last three columns)

ChatGPT

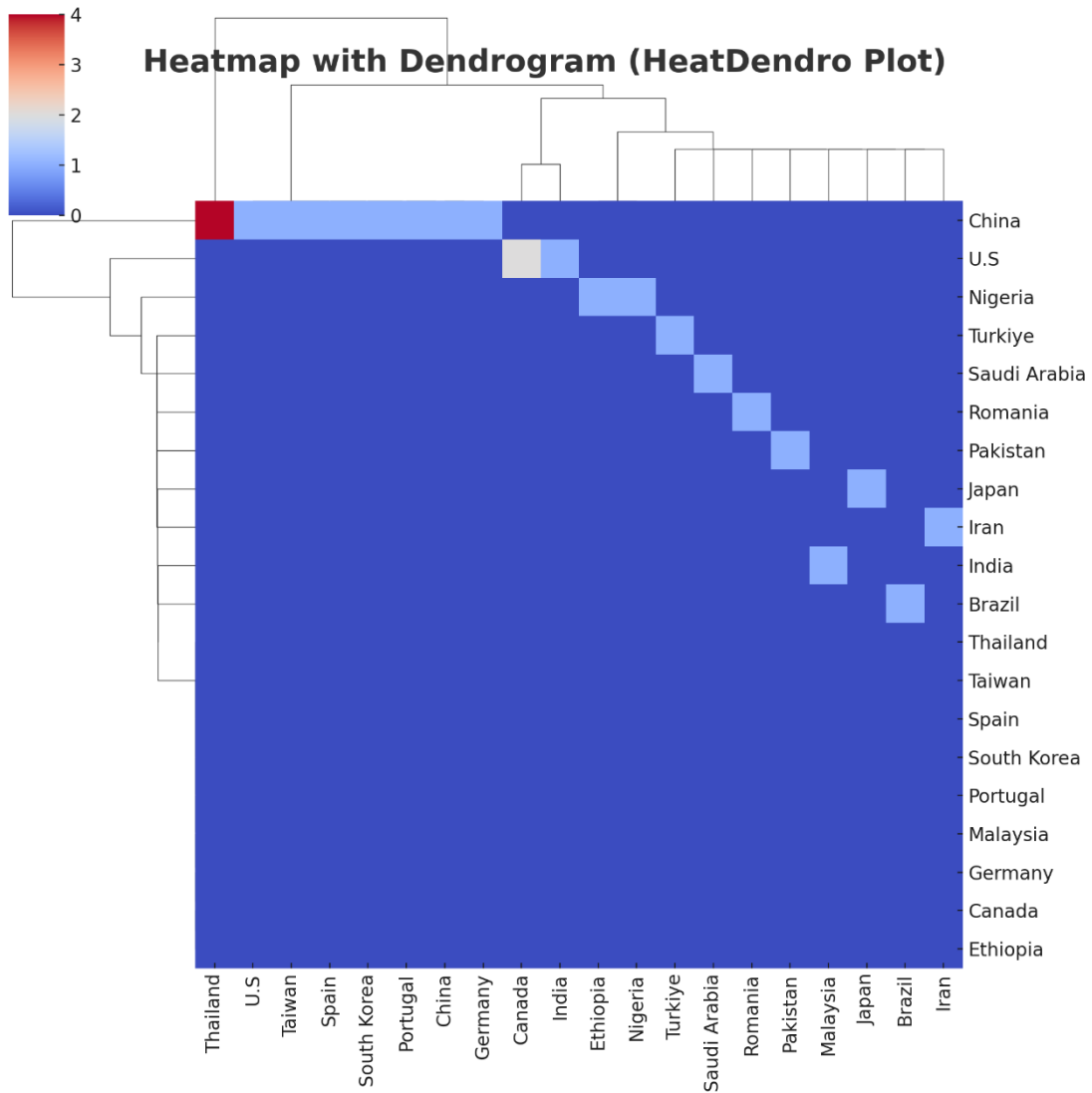
Certainly! To create a Heatmap with Dendrogram (also known as a HeatDendro plot) similar to the one you uploaded, I will use the relation data in the last three columns of the previously uploaded file. In this HeatDendro plot:

- The **rows** and **columns** of the heatmap represent entities (nodes), which are based on the relations in the data.
- The **color intensity** in the heatmap represents the strength or weight of the connections between entities, as indicated by the values in the data.
- The **dendrograms** on the top and left side of the heatmap represent hierarchical clustering of entities based on their connections, helping to visually group similar entities together.

Let's start drawing the HeatDendro plot based on this data.

Finished working

Show work

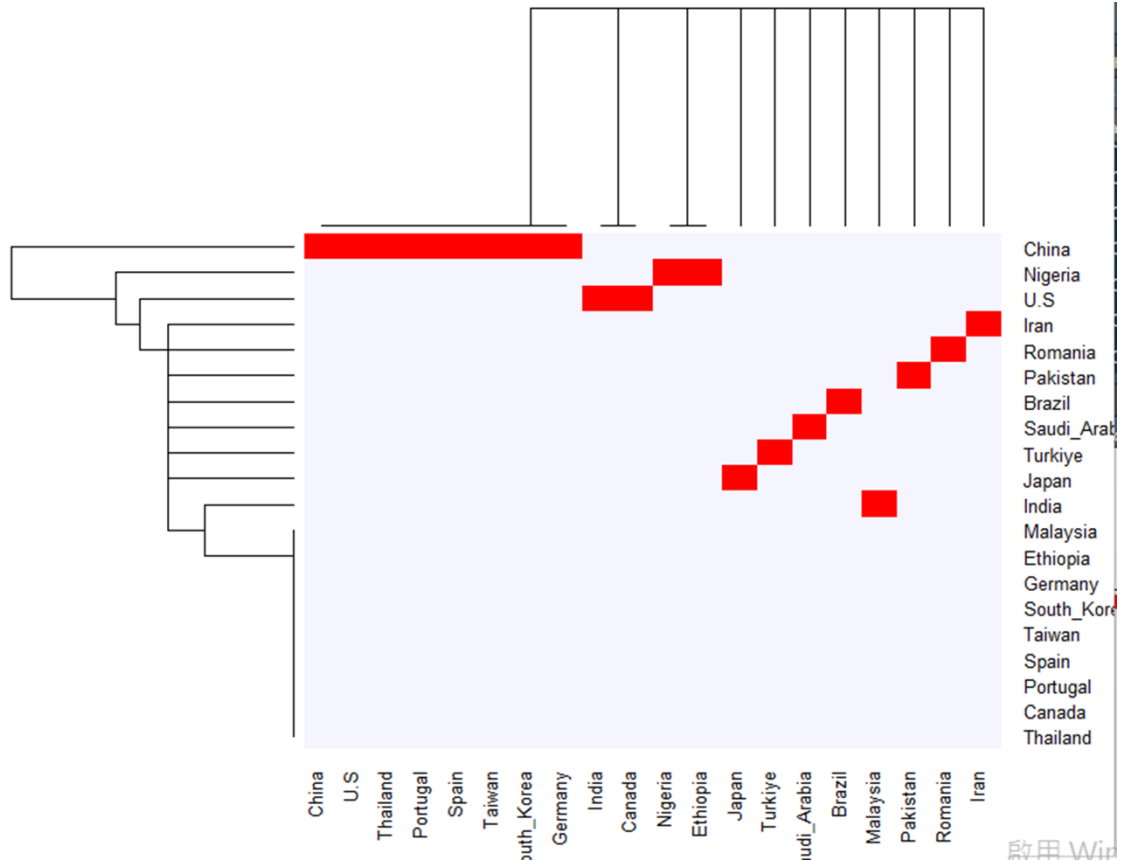


Here is the Heatmap with Dendrogram (also known as a HeatDendro plot) based on the relation data in the last three columns of the previously uploaded file:

- The **rows** and **columns** of the heatmap represent entities (nodes), and the labels indicate the names of these entities.
- The **color intensity** in the heatmap represents the strength or weight of the connections between entities. Warmer colors (towards red) indicate higher weights, and cooler colors (towards blue) indicate lower weights.

- The **dendrograms** on the top and left side of the heatmap represent the hierarchical clustering of entities based on their connections. This clustering helps to visually group similar entities together.

I hope this visualization meets your expectations. Please let me know if you would like any further adjustments or if there are specific features you would like to add to this plot!



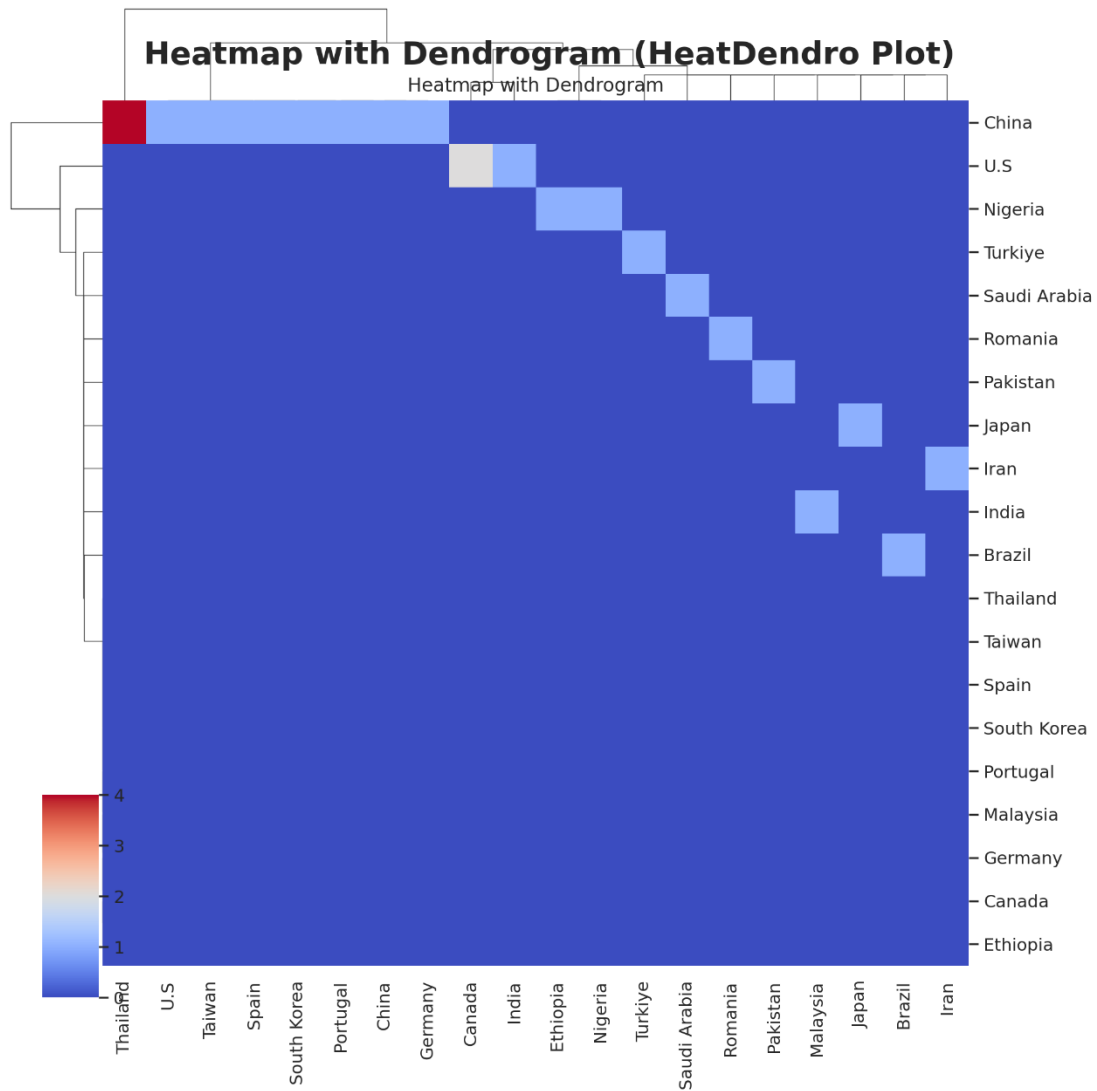
But, the background is still blue instead of white as I requested

ChatGPT

My apologies for the oversight. I will make sure to set the background color of the heatmap to white as you requested. Let me correct this and show you the updated HeatDendro plot with a white background.

Finished working

Show work



Here is the updated Heatmap with Dendrogram (HeatDendro plot) with a white background, as per your request:

- The **rows** and **columns** of the heatmap represent entities (nodes), and the labels indicate the names of these entities.
- The **color intensity** in the heatmap represents the strength or weight of the connections between entities. Warmer colors (towards red) indicate higher weights, and cooler colors (towards blue) indicate lower weights.
- The **dendrograms** on the top and left side of the heatmap represent the hierarchical clustering of entities based on their connections. This clustering helps to visually group similar entities together.

The background of the entire HeatDendro plot, including the dendrogram and heatmap, is now set to white, making the visualization cleaner and easier to read.

Please let me know if this meets your expectations, and if there are any further adjustments or specific features you would like to add to this plot!

But, the background is still blue instead of white i requested

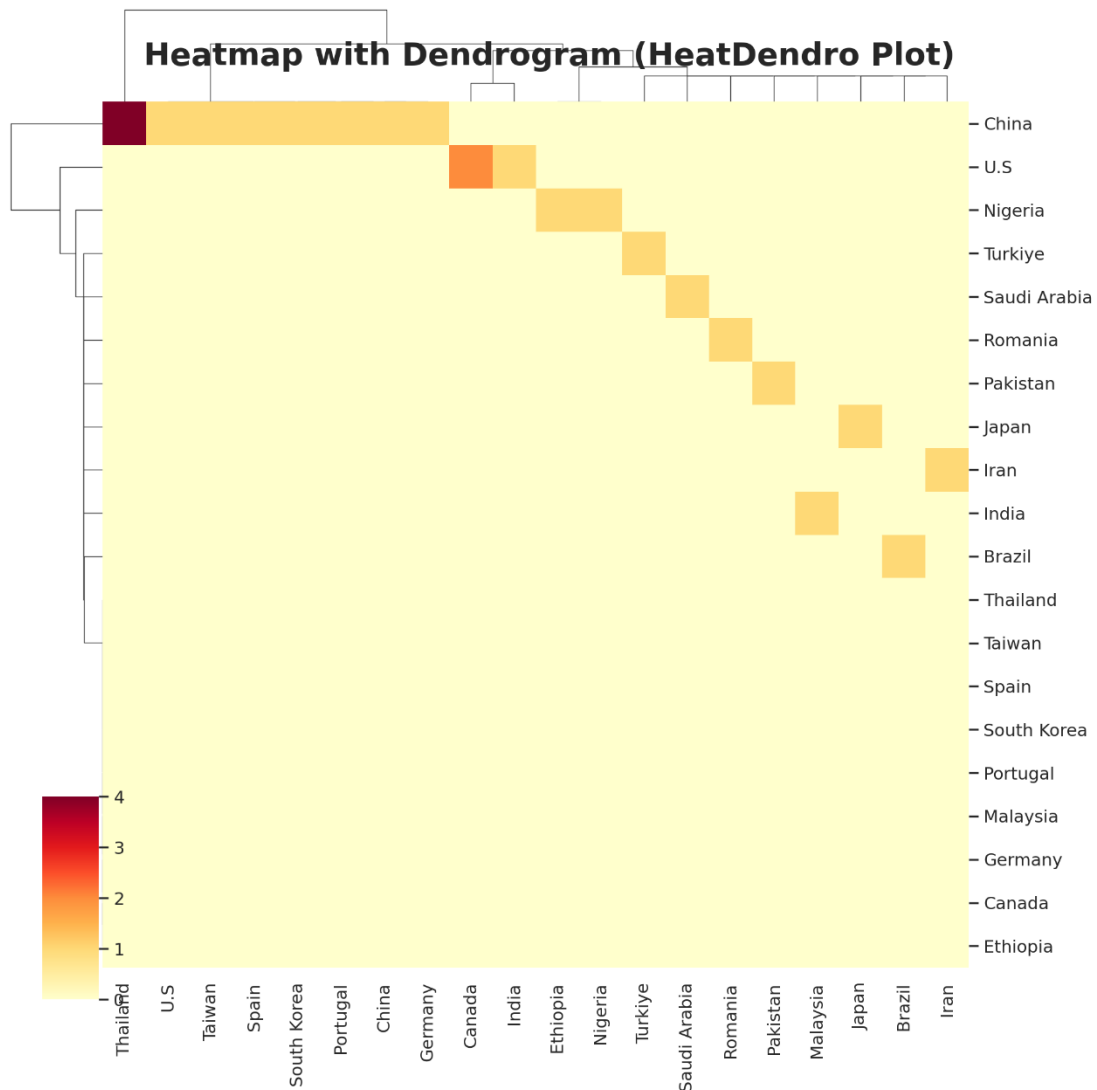
ChatGPT

My apologies for the oversight. I misunderstood your request earlier. I will make sure to set the background of the cells in the heatmap to white, which I believe is what you are referring to.

Let me correct this and show you the updated HeatDendro plot with a white background for the cells.

Finished working

Show work



Here is the updated Heatmap with Dendrogram (HeatDendro plot) with a white background for the cells, as per your request:

- The **rows** and **columns** of the heatmap represent entities (nodes), and the labels indicate the names of these entities.
- The **color intensity** in the heatmap represents the strength or weight of the connections between entities. Warmer colors (towards red) indicate higher weights, while white indicates lower weights or absence of a connection.
- The **dendrograms** on the top and left side of the heatmap represent the hierarchical clustering of entities based on their connections. This clustering helps to visually group similar entities together.

The background of the cells in the HeatDendro plot is now set to white, making the visualization cleaner and easier to read.

Please let me know if this meets your expectations, and if there are any further adjustments or specific features you would like to add to this plot!

But, the background is still yellow instead of white I expected

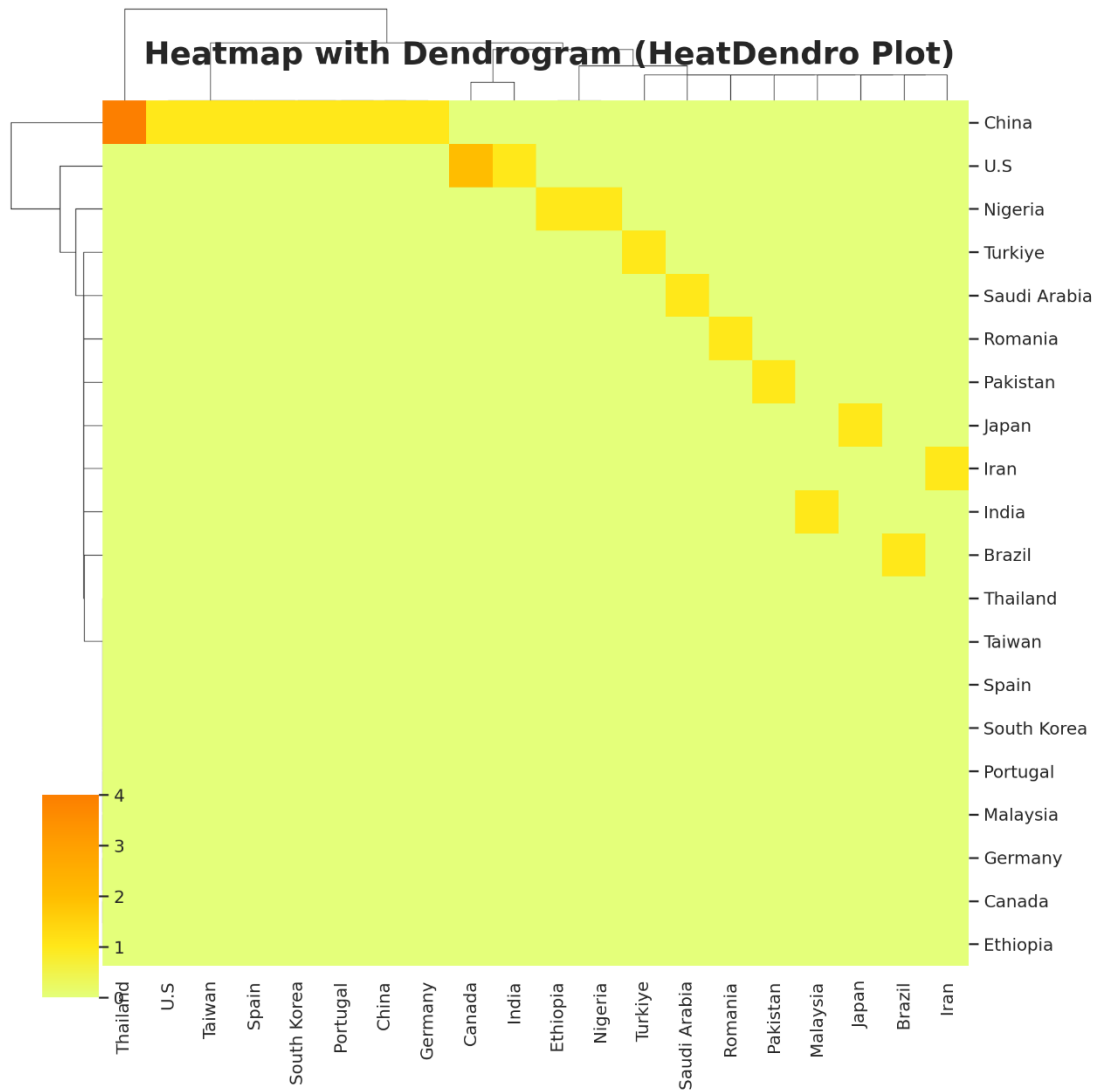
ChatGPT

My apologies for the confusion earlier. I understand now that you are referring to the background color of the cells in the heatmap itself, not the overall plot background.

I will set the background color of the cells in the heatmap to white for the lowest values, and maintain the color gradient for higher values. Let me make this correction and show you the updated HeatDendro plot.

Finished working

Show work



Here is the updated Heatmap with Dendrogram (HeatDendro plot) with a white background for the cells, as per your request:

- The **rows** and **columns** of the heatmap represent entities (nodes), and the labels indicate the names of these entities.
- The **color intensity** in the heatmap represents the strength or weight of the connections between entities. The color transitions from white (for lower weights or absence of a connection) to warmer colors (for higher weights).
- The **dendrograms** on the top and left side of the heatmap represent the hierarchical clustering of entities based on their connections. This clustering helps to visually group similar entities together.

The background of the cells in the HeatDendro plot is now set to white, making the visualization cleaner and easier to read.

Please let me know if this meets your expectations, and if there are any further adjustments or specific features you would like to add to this plot!

But, the background is light green instead of white I expected

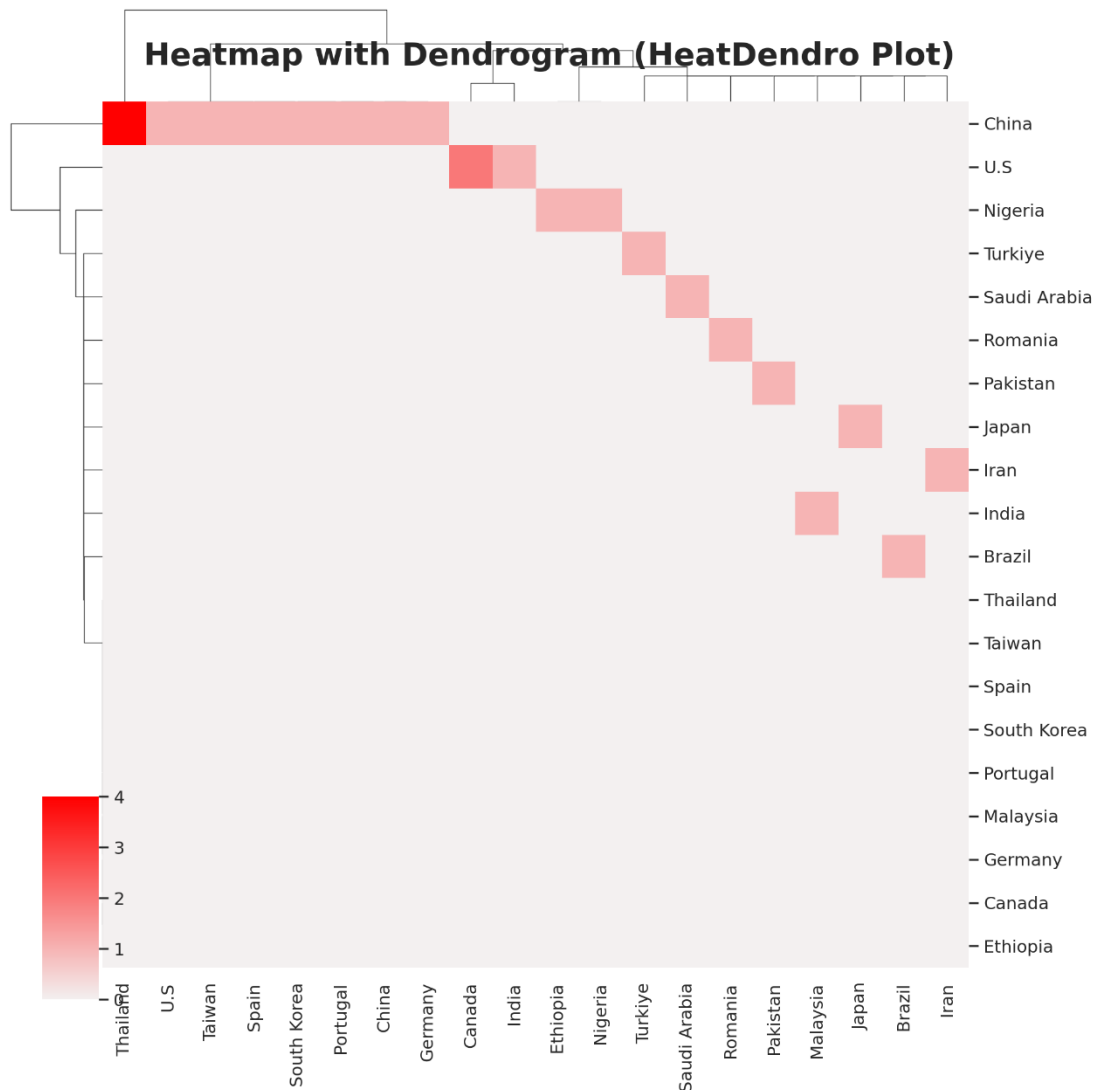
ChatGPT

I apologize for the misunderstandings earlier. I will modify the colormap to ensure that the background color of the cells in the heatmap is explicitly set to white for the lowest values, while maintaining a color gradient for higher values.

Let's make this correction and show the updated HeatDendro plot with a white background for the cells.

Finished working

Show work



Here is the updated Heatmap with Dendrogram (HeatDendro plot) with a white background for the cells, as per your request:

60A7B.D1:3 column to 3 column data

60A7B.D1:3 to 4 for Network in SNA

Submit

Tips for R The number of Clusers is set at (2,targe, value) = 2, for example
Copy & Paste the Code in R to Rstudio



Select text

```

Leader,follower,cluster,WCD,2
China,China,1,1134,1
South_Korea,South_Korea,1,181,1
Portugal,Portugal,1,116,1
Japan,Japan,2,95,2
Turkiye,Turkiye,3,66,3
Taiwan,Taiwan,1,62,1
U.S,U.S,7,44,7
Saudi_Arabia,Saudi_Arabia,4,26,4
Brazil,Brazil,5,17,5
Germany,Germany,1,11,1
India,India,6,9,6
Canada,Canada,7,9,7
Malaysia,Malaysia,6,9,6
Spain,Spain,1,8,1
Nigeria,Nigeria,9,8,9

```

1.Refresh

2.Paste to input box

3.Click on Me

Next to:Authorcorrence9A

(Clear/Type/Submit to see data format)

Clear

```

Source,Target,WCD
Brazil,Brazil,17
Canada,Canada,9
Canada,U.S,2
China,China,1134
China,Portugal,1
China,South_Korea,1
China,Thailand,4
Ethiopia,Ethiopia,5
Ethiopia,Nigeria,1
China,Germany,1
Germany,Germany,11
India,India,9
India,U.S,1
Iran,Iran,6

```

plots in R

Chinese notes

Submit for Authorcorrence9A

Click on 1,2,3


```

library(igraph)
data <-
data.frame(Leader=c("China","Canada","China","China","Ethiopia","China","India","I
ndia","China","China","China","Portugal")
,follower=c("Thailand","U.S","South_Korea","Portugal","Nigeria","Germany","U.S","
Malaysia","Spain","Taiwan","U.S","U.S")
,cluster=c(1,2,1,1,3,1,2,2,1,1,1,1)
,WCD=c(4,2,1,1,1,1,1,1,1,1,1,1)
,clusterB2=c(1,2,1,1,3,1,2,2,1,1,2,2)
)
#WCD=c(1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1)Portugal
U.S
China
Taiwan
Spain
India
Malaysia
Germany
Ethiopia
Nigeria
South_Korea
Canada
Thailand

nodes <-
data.frame(name=c("Portugal","U.S","China","Taiwan","Spain","India","Malaysia","G
ermany","Ethiopia","Nigeria","South_Korea","Canada","Thailand"),
carac=c("1","2","1","1","1","2","2","1","3","3","1","2","1"),
value=c(2,5,10,1,1,2,1,1,1,1,1,2,4),
value2=c(2,5,10,1,1,2,1,1,1,1,1,2,4))

#nodes$value2      used for computing original counts in textabc2 set

# data<-as.matrix.data.frame(data)
# df.g <- graph.data.frame(d = data[,1:2], directed = FALSE)
# plot(df.g, vertex.label = V(df.g)$name,vertex.color = "green")

```



```

# Turn it into igraph object
network <- graph_from_data_frame(d=data, vertices=nodes, directed=F)

# Make a palette of 3 colors
library(RColorBrewer)
coul <- brewer.pal(3, "Set1")

# Create a vector of color
my_color <- coul[as.numeric(as.factor(V(network)$carac))]
deg <- degree(network, mode="all")
V(network)$size <- nodes$value2*3 #deg*3
# We could also use the audience size value:
#V(network)$size <- nodes$value*1.0
# nodes are sized by value of TWCD
# https://kateto.net/netscix2016.html
# Make the plot
# plot(network, vertex.color=my_color)

# Add a legend
#legend("bottomleft", legend=levels(as.factor(V(network)$carac)) , col = coul , bty
= #"n", pch=20 , pt.cex = 3, cex = 1.5, text.col=coul , horiz = FALSE, inset = c(0.1, 0.1))
#plot(network, vertex.color=my_color, edge.width=E(network)$WCD*2 )
legend("bottomleft", legend=levels(as.factor(V(network)$carac)) , col = coul , bty =
"n", pch=20 , pt.cex = 3, cex = 1.5, text.col=coul , horiz = FALSE, inset = c(0.1, 0.1))
plot(network, vertex.color=my_color,
edge.width=E(network)$WCD*1/20,vertex.label.dist=1)
#Adjust : E(network)$value*1/20 and V(network)$size <- deg*3 for edge and bubble
sizes
#legend(x=-1.1, y=-1.1, c("Newspaper", "Television", "Online News"), pch=21,
#col="#777777", pt.bg=my_color, pt.cex=2.5, bty="n", ncol=1)
#https://oak.ucc.nau.edu//yc424/\_book/data-to-network.html
#plot(network, vertex.shape="none", vertex.label=V(net)$media,
# vertex.label.font=2, vertex.label.color="gray40",
# vertex.label.cex=.7, edge.color="gray85")

#edge.start <- ends(network, es=E(network), names=F)[,1]

```

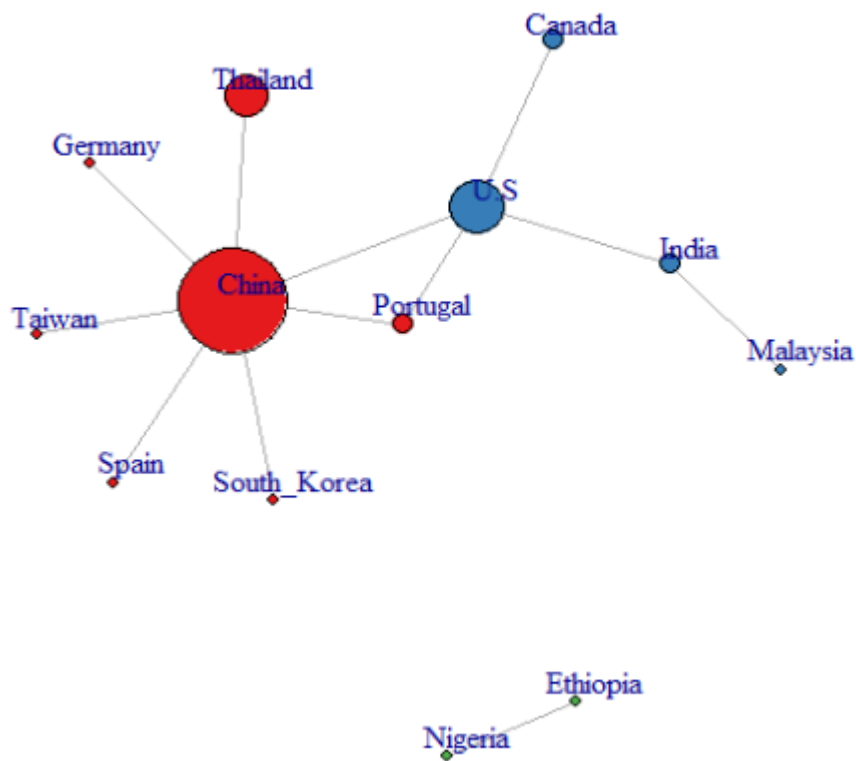
```
#edge.col <- V(network)$color[edge.start]
```

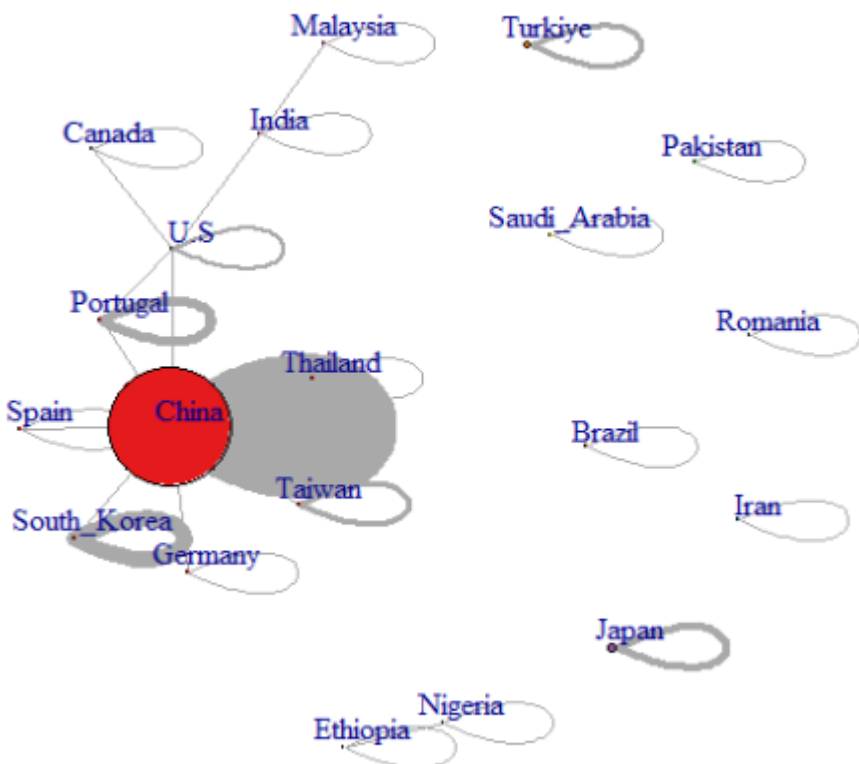
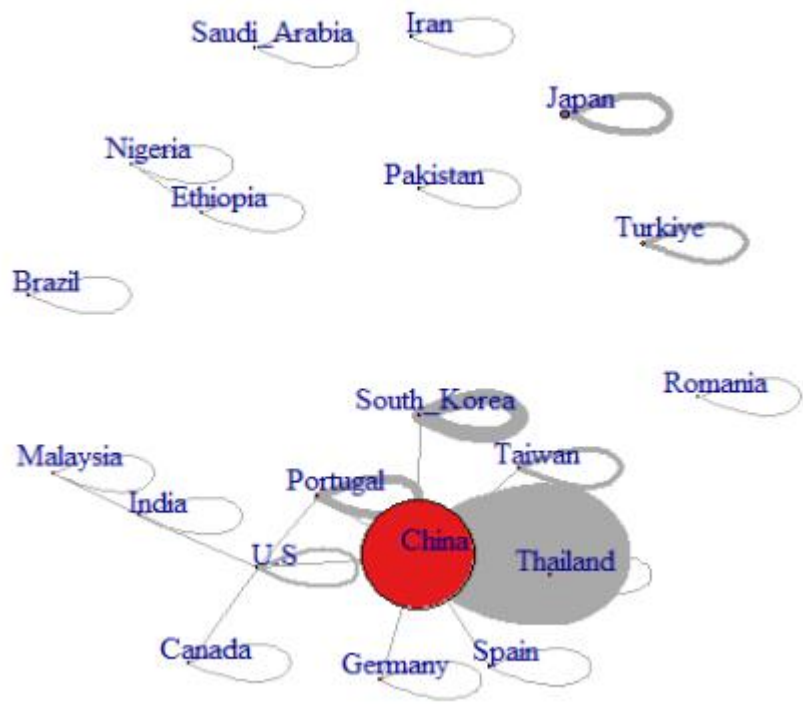
```
#plot(network, edge.color=edge.col, edge.curved=.1)
```

```
#ceb<-cluster_edge_betweenness(network)
```

```
#dendPlot(ceb,mode="hclust")
```

```
#plot(ceb, network)
```

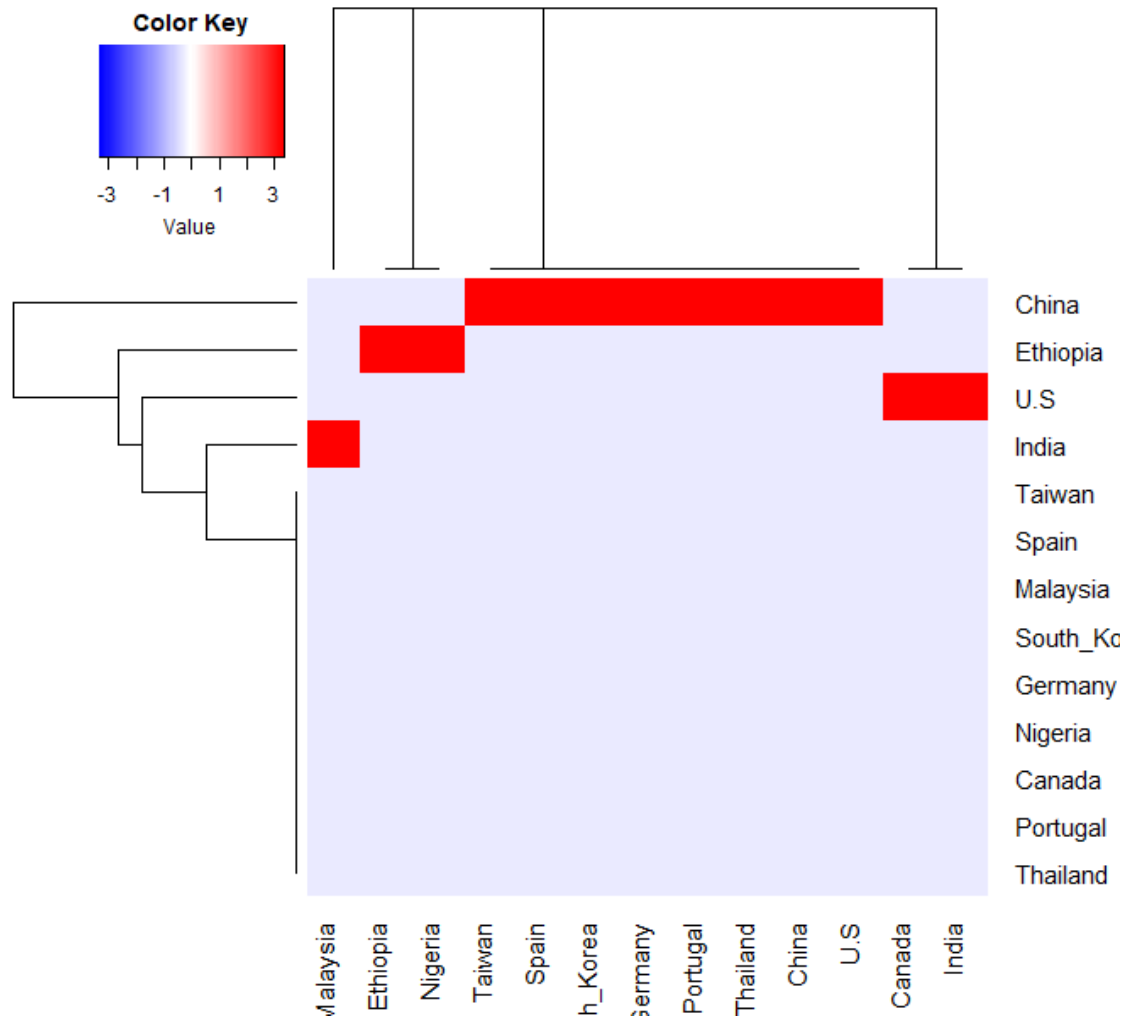




Source,Target,WCD

Canada,U.S,2
China,Portugal,1
China,South_Korea,1
China,Thailand,4
Ethiopia,Nigeria,1
China,Germany,1
India,U.S,1
India,Malaysia,1
China,Spain,1
China,Taiwan,1
China,U.S,1
Portugal,U.S,1

60A4.*CircleBar. to #15 CircleBar
60A5.*CirclePacking. to #34 CirclePacking
60A6.D4:*HeatDendro. to #52 HeatDendr
60A7.D55:*Network: next to #46
60A7.D5:*Network: next to #40A
Uti: To Pack by 60A7 then to 40A



```

# Libraries
library("RColorBrewer")
library(heatmap)
library(heatmapfit)
library(heatmap.2)
library(pheatmap)
library(d3heatmap)
library("gplots")

data <-
data.frame(model=c("China","U.S","Thailand","Portugal","Canada","India","Ethiopia"
,"Nigeria","Germany","South_Korea","Malaysia","Spain","Taiwan")
,China=c(1,0,0,0,0,0,0,0,0,0,0,0,0)

```

```
,U.S=c(1,0,0,0,0,0,0,0,0,0,0)
,Thailand=c(4,0,0,0,0,0,0,0,0,0,0)
,Portugal=c(1,0,0,0,0,0,0,0,0,0,0)
,Canada=c(0,2,0,0,0,0,0,0,0,0,0)
,India=c(0,1,0,0,0,0,0,0,0,0,0)
,Ethiopia=c(0,0,0,0,0,1,0,0,0,0,0)
,Nigeria=c(0,0,0,0,0,1,0,0,0,0,0)
,Germany=c(1,0,0,0,0,0,0,0,0,0,0)
,South_Korea=c(1,0,0,0,0,0,0,0,0,0,0)
,Malaysia=c(0,0,0,0,1,0,0,0,0,0,0)
,Spain=c(1,0,0,0,0,0,0,0,0,0,0)
,Taiwan=c(1,0,0,0,0,0,0,0,0,0,0)
)
```

```
dev.off()
```

```
df<-scale(data[2:14])
```

```
rownames(df) <- c(data$model)
```

```
library("RColorBrewer")
```

```
#col <- colorRampPalette(brewer.pal(10, "RdYlBu"))(256)
```

```
col<- colorRampPalette(c("yellow", "white", "blue"))(256)
```

```
heatmap(df, scale = "none", col = col,RowSideColors
=c(rep("blue",4),rep("pink",9)),
```

```
ColSideColors = c(rep("purple", 5), rep("orange", 8)))
```

```
# none->row or column
```

```
heatmap.2(df, scale = "none", col = bluered(100),
```

```
trace = "none", density.info = "none")
```

```
# Set color palette
```

```
my_palette <- colorRampPalette(c("white", "white", "red"))(n =169)
```

```
#library("pheatmap")
```

```
# Create heatmap
```

```
#pheatmap(df, cutree_rows = 4, color = my_palette)
```

```
#pheatmap(df, border_color = "white",
```

```
#cellwidth = 9, cellheight = 9)
```

```
#pheatmap(df, display_numbers = TRUE,
```

```
# number_color = "blue", number_format = "%.1e") #default "%.2f"
```

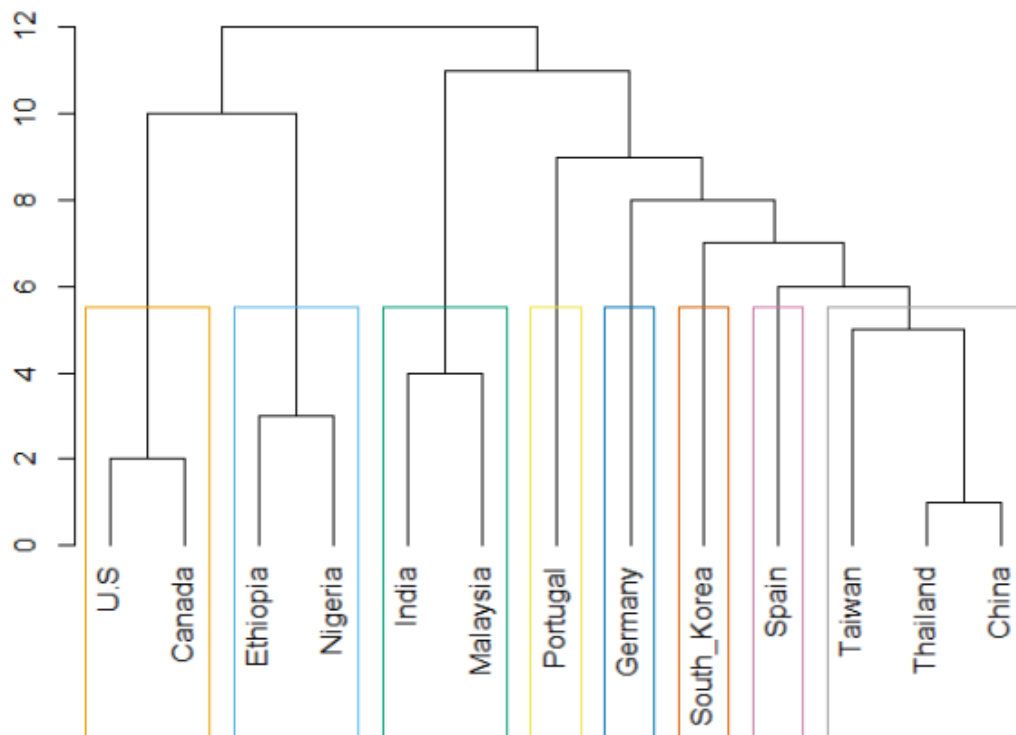
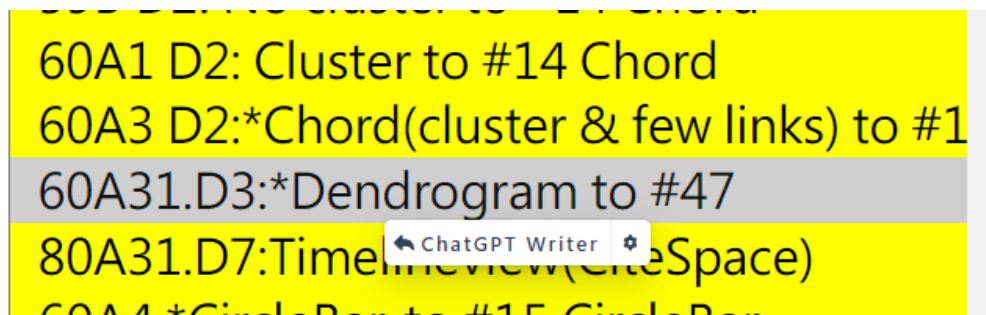
#number_format = "%d": display numbers as integers
#number_format = "%.2f": display numbers as fixed-point decimals #with two decimal places
#number_format = "%.0f%%": display numbers as percentages with #
#number_format = "%.1e": display numbers in scientific notation #with one decimal place
#number_format = "\$%.2f": display numbers as currency values #with two decimal places and a dollar sign

59B D2: No cluster to #14 Chord
60A1 D2: Cluster to #14 Chord
60A3 D2:*Chord(cluster & few links) to #1
60A31.D3:*Dendrogram to #47
80A31.D7:Timelineview(CiteSpace)
60A4.*CircleBar. to #15 CircleBar


```
dimnames(m) <- list(have = haircolors, prefer = haircolors)
groupColors <- c("#96ab03", "#96ab03", "#96ab03", "#96ab03", "#96ab03",
"#96ab03", "#96ab03", "#6495ED", "#6495ED", "#6495ED", "#6495ED", "#FF7F50",
"#FF7F50")
```

```
p <- chorddiag(m, groupColors = groupColors, groupnamePadding = 20)
```

```
p
```



```
#library
library(igraph)
```



```

#legend(x=-1.5, y=-1.1, c("Newspaper", "Television", "Online News"), pch=21,
#edge.start <- ends(net, es=E(net), names=F)[,1]
#edge.col <- V(net)$color[edge.start]

#plot(net, edge.color=edge.col, edge.curved=.1)
# col="#777777", pt.bg=colrs, pt.cex=2, cex=.8, bty="n", ncol=1)
# https://kateto.net/netscix2016.html
#net <- graph_from_data_frame(d=links, vertices=nodes, directed=T)
#net2 <- graph_from_incidence_matrix(links2)
# plot(net2, vertex.shape="none", vertex.label=nodes2$media,

# vertex.label.color=V(net2)$color, vertex.label.font=2.5,

# vertex.label.cex=.6, edge.color="gray70", edge.width=2)
#deg <- degree(net, mode="all")

#plot(net, vertex.size=deg*3)
#hs <- hub_score(net, weights=NA)$vector

#as <- authority_score(net, weights=NA)$vector

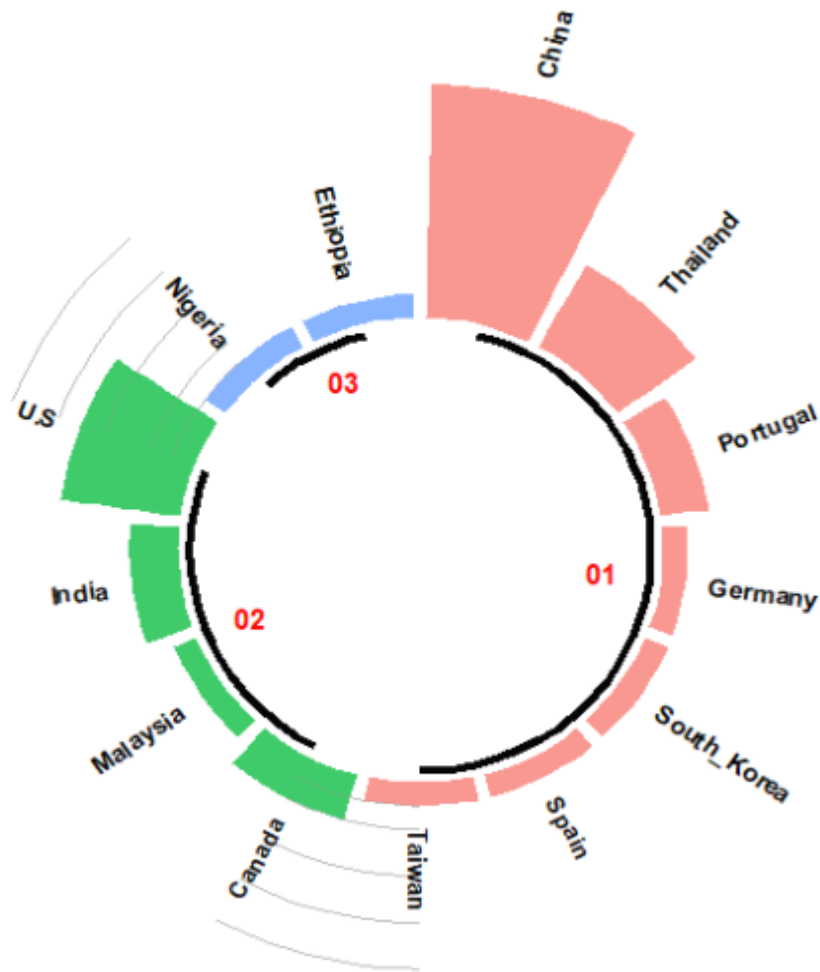
ceb <- cluster_edge_betweenness(network)
dendPlot(ceb, mode="hclust")
#Community detection based on edge betweenness (Newman-Girvan)
#High-betweenness edges are removed sequentially (recalculating at each step) #and
the best partitioning of the network is selected.

```

```

60A1.D7: *CircleView. to #15 CircleView
60A4.*CircleBar. to #15 CircleBar
60A5.*CirclePacking. to #34 CirclePacking
60A6.D4:*HeatDendro. to #52 HeatDendro

```



```
library(tidyverse)
# Create dataset
data <-
data.frame(individual=c("China","Thailand","Portugal","Germany","South_Korea","S
pain","Taiwan","Canada","Malaysia","India","U.S","Nigeria","Ethiopia")
,group=c("01","01","01","01","01","01","01","02","02","02","02","03","03")
,value=c(100,40,20,10,10,10,20,10,20,10,20,50,10,10)
)
```

```
empty_bar <- 0
to_add <- data.frame( matrix(NA, empty_bar*nlevels(data$group), ncol(data)) )
colnames(to_add) <- colnames(data)
to_add$group <- rep(levels(data$group), each=empty_bar)
data <- rbind(data, to_add)
data <- data %>% arrange(group)
data$id <- seq(1, nrow(data))
```

```

# Get the name and the y position of each label
label_data <- data
number_of_bar <- nrow(label_data)
angle <- 90 - 360 * (label_data$id-0.5) / number_of_bar      # I subtract 0.5
because the letter must have the angle of the center of the bars. Not extreme
right(1) or extreme left (0)
label_data$hjust <- ifelse( angle < -90, 1, 0)
label_data$angle <- ifelse(angle < -90, angle+180, angle)

# prepare a data frame for base lines
base_data <- data %>%
  group_by(group) %>%
  summarize(start=min(id), end=max(id) - empty_bar) %>%
  rowwise() %>%
  mutate(title=mean(c(start, end)))

# prepare a data frame for grid (scales)
grid_data <- base_data
grid_data$end <- grid_data$end[ c( nrow(grid_data), 1:nrow(grid_data)-1)] + 1
grid_data$start <- grid_data$start - 1
grid_data <- grid_data[-1,]

# Make the plot
p <- ggplot(data, aes(x=as.factor(id), y=value, fill=group)) +      # Note that id is a
factor. If x is numeric, there is some space between the first bar

  geom_bar(aes(x=as.factor(id), y=value, fill=group), stat="identity", alpha=0.5) +

  # Add a val=100/75/50/25 lines. I do it at the beginning to make sur barplots are
OVER it.
  geom_segment(data=grid_data, aes(x = end, y = 80, xend = start, yend = 80),
colour = "grey", alpha=1, size=0.3 , inherit.aes = FALSE ) +
  geom_segment(data=grid_data, aes(x = end, y = 60, xend = start, yend = 60),
colour = "grey", alpha=1, size=0.3 , inherit.aes = FALSE ) +
  geom_segment(data=grid_data, aes(x = end, y = 40, xend = start, yend = 40),
colour = "grey", alpha=1, size=0.3 , inherit.aes = FALSE ) +
  geom_segment(data=grid_data, aes(x = end, y = 20, xend = start, yend = 20),

```

```

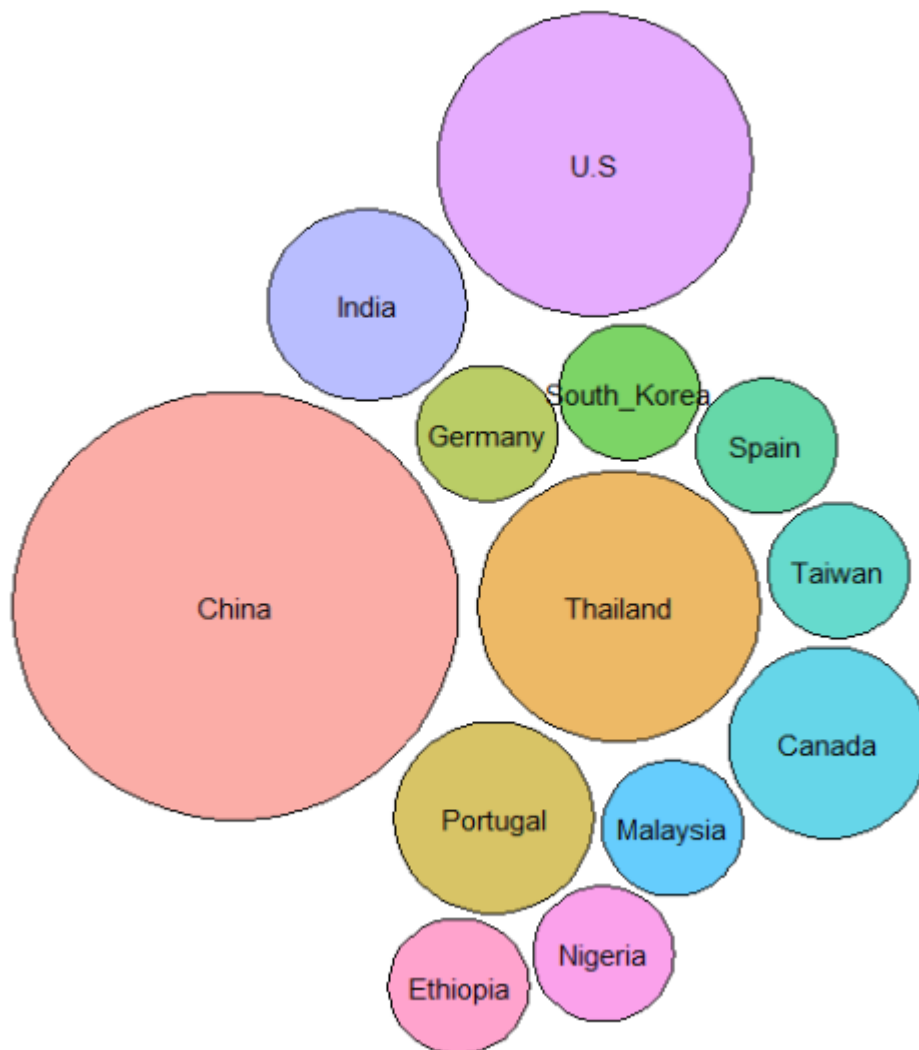
colour = "grey", alpha=1, size=0.3 , inherit.aes = FALSE ) +
  geom_segment(data=grid_data, aes(x = end, y = 10, xend = start, yend = 10),
colour = "grey", alpha=1, size=0.3 , inherit.aes = FALSE ) +
  # Add text showing the value of each 100/75/50/25 lines
#annotate("text", x = rep(max(data$id),4), y = c(20, 40, 60, #80), label =
c("20","40","60","80") , color="red", size=3 , #angle=0, fontface="bold" , hjust=1) +

  geom_bar(aes(x=as.factor(id), y=value, fill=group), stat="identity", alpha=0.5) +
  ylim(-100,120) +
  theme_minimal() +
  theme(
    legend.position = "none",
    axis.text = element_blank(),
    axis.title = element_blank(),
    panel.grid = element_blank(),
    plot.margin = unit(rep(-1,4), "cm")
  ) +
  coord_polar() +
  geom_text(data=label_data, aes(x=id, y=value+10, label=individual, hjust=hjust),
color="black", fontface="bold",alpha=0.98, size=3.5, angle= label_data$angle,
inherit.aes = FALSE ) +

  # Add base line information
  geom_segment(data=base_data, aes(x = start, y = -5, xend = end, yend = -5),
colour = "black", alpha=0.98 , size=1.6 , inherit.aes = FALSE ) +
  geom_text(data=base_data, aes(x = title, y = -18, label=group), hjust=c(1,0,0), colour
= "red", alpha=0.98, size=4, fontface="bold", inherit.aes = FALSE)
p

```

```
80A31.D7:Timelineview(CiteSpace)
60A4.*CircleBar. to #15 CircleBar
60A5.*CirclePacking. to #34 CirclePacking
60A6.D4:*HeatDendro. to #52 HeatDendr
60A7.D55:*Network: next to #46
60A7.D5:*Network: next to #40A
```



```
# libraries
library(packcircles)
library(ggplot2)
library(viridis)
data <-
data.frame(individual=c("China","Thailand","Portugal","Germany","South_Korea","S
```

```

pain","Taiwan","Canada","Malaysia","India","U.S","Nigeria","Ethiopia")
,value=c(10,4,2,1,1,1,1,2,1,2,5,1,1)
)

# Create data

# Generate the layout. sizetype can be area or radius, following your preference on
what to be proportional to value.
packing <- circleProgressiveLayout(data$value, sizetype='area')
packing$radius <- 0.95*packing$radius
dat.gg <- circleLayoutVertices(packing, npoints=50)
data <- cbind(data, packing)
# Basic color customization
ggplot() +
  geom_polygon(data = dat.gg, aes(x, y, group = id, fill=as.factor(id)), colour =
"black", alpha = 0.6) +

  geom_text(data = data, aes(x, y, size=value, label = individual)) +
  scale_size_continuous(range = c(1,4)) +
  theme_void() +
  theme(legend.position="none") +
  coord_equal()

```

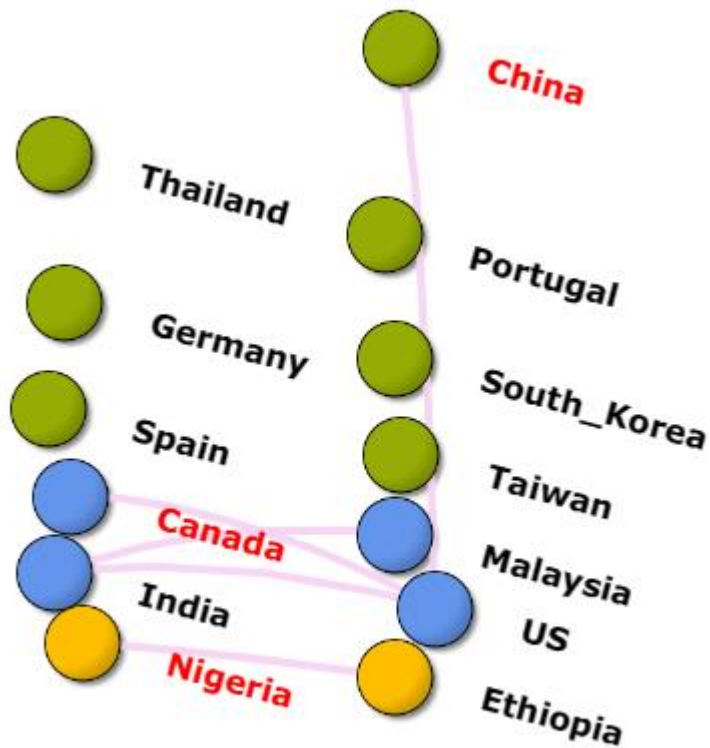
60A9.D6:Google Maps(Pajek in 40A)

60A8: *Details about cluster process

60A88: *Details without self connections

60A7A.D5:Data from 60A8 to pajek for Go

60A7A.D5:Data from 60A8 to pajek for Go



HTML file

```

<!DOCTYPE html PUBLIC "-//W3C//DTD XHTML 1.0 Strict//EN"
"http://www.w3.org/TR/xhtml1/DTD/xhtml1-strict.dtd">

<html>
<html xmlns="http://www.w3.org/1999/xhtml">

  <head>
<meta http-equiv="Content-Type" content="text/html; charset=big5" />

  <meta http-equiv="Content-Type" content="text/html; charset=utf-8" />

<title>PaperABC_GoogleMap</title>
<meta name="viewport" content="width=device-width; initial-scale=1.0">

<script src="../autoadjust/js/jquery-1.7.1.min.js"></script>

```

```

<script src="../../autoadjust/js/script.js"></script>
<script src="../../autoadjust/js/forms.js"></script>
<script src="../../autoadjust/js/superfish.js"></script>

<script src="../../autoadjust/js/jquery.responsivemenu.js"></script>

<script src="../../autoadjust/js/FF-cash.js"></script>

<script>
$(function(){
    $('#contact-form').forms({ownerEmail:'#'});
});
</script>

<!DOCTYPE html>
<!--[if lt IE 7 ]> <html lang="zh-tw" class="no-js ie6"> <![endif]-->

<!--[if IE 7 ]>    <html lang="zh-tw" class="no-js ie7"> <![endif]-->

<!--[if IE 8 ]>    <html lang="zh-tw" class="no-js ie8"> <![endif]-->

<!--[if IE 9 ]>    <html lang="zh-tw" class="no-js ie9"> <![endif]-->

<!--[if (gt IE 9) | !(IE)]><!-->
<html lang="zh-tw" class="no-js"> <!--<![endif]-->

        <title>PaperABC(Smile Chien)</title>
        <base target="_top"></base>
<meta name=Sgeo.positionS content="0;0" />
<meta name="ICBM" content="0,0" />
<meta http-equiv="Content-Type" content="text/html; charset=UTF-8"
/>

<style>
body {
    background-color:#E7E9EB;
}
#myDIV {

```

```

        height:300px;
        background-color:#FFFFFF;
    }
    #blueDIV {
        position:relative;
        width:100px;
        padding:20px;
        transform: rotate(15deg);
    }
    #chienDIV {
        .translate-rotate {
            transform: translateX(180px) rotate(45deg);
        }
    }
</style>
</head>
<body onload="GUnload();" style="margin:0px;">
<script>
google_api_key = 'AlzaSyBPTqFXDLNP2Og71yuXWZtNcrDiRC-BPsc'; // Your project's
Google Maps API key goes here (https://code.google.com/apis/console)
    language_code = "";
    document.writeln('<script type="text/javascript"
src="https://maps.googleapis.com/maps/api/js?libraries=geometry'+((self.google_ap
i_key&&document.location.toString().indexOf('file://')!=0)?'&key='+google_api_key:''
)+'"><'+'/script>');
</script>

<div id="gmap_div" style="width:700px; height:522px; margin:0px;
margin-right:12px; background-color:#F0F0F0; float:left; overflow:hidden;">

    <p align="center" style="font:10px Arial;">This map was
created using <a target="_blank" href="http://www.gpsvisualizer.com/">GPS
Visualizer</a>'s do-it-yourself geographic utilities.<br /><br />Please wait while the
map data loads...</p>
</div>

```

```

<script type="text/javascript">
  /* Global variables used by the GPS Visualizer functions
(20150703144805): */
  gv_options = {};
  // basic map parameters:
  gv_options.center = [0,0]; // [latitude,longitude] - be sure to keep the
square brackets
  gv_options.zoom = 1;
  gv_options.map_type = 'US_NATIONAL_ATLAS'; // popular map_type
choices are 'GV_STREET', 'GV_SATELLITE', 'GV_HYBRID', 'GV_TERRAIN',
'GV_TOPO_US', 'GV_TOPO_WORLD', 'GV_OSM'
  gv_options.map_opacity = 0.00; // number from 0 to 1

  gv_options.full_screen = true; // true|false: should the map fill the entire
page (or frame)?
  gv_options.width = 700; // width of the map, in pixels

  gv_options.height = 700; // height of the map, in pixels

  gv_options.map_div = 'gmap_div'; // the name of the HTML "div" tag
containing the map itself; usually 'gmap_div'
  gv_options.doubleclick_zoom = true; // true|false: zoom in when mouse
is double-clicked?
  gv_options.doubleclick_center = true; // true|false: re-center the map on
the point that was double-clicked?
  gv_options.mousewheel_zoom = true; // true|false; or 'reverse' for
down=in and up=out
  gv_options.autozoom_adjustment = 0;
  gv_options.centering_options = { 'open_info_window':true,
'partial_match':true, 'center_key':'center', 'default_zoom':null } // URL-based
centering (e.g., ?center=name_of_marker&zoom=14)
  gv_options.tilt = false; // true|false: allow Google to show 45-degree
tilted aerial imagery?
  gv_options.street_view = false; // true|false: allow Google Street View on
the map
  gv_options.animated_zoom = false; // true|false: may or may not work

```

properly

`gv_options.disable_google_pois = false; // true|false: if you disable clickable POIs, you also lose the labels on parks, airports, etc.`

`// widgets on the map:`

`gv_options.zoom_control = 'small'; // 'large'|'small'|'none'`

`gv_options.recenter_button = true; // true|false: is there a 'double-click to recenter' option in the zoom control?`

`gv_options.scale_control = true; // true|false`

`gv_options.center_coordinates = true; // true|false: show a "center coordinates" box and crosshair?`

`gv_options.mouse_coordinates = false; // true|false: show a "mouse coordinates" box?`

`gv_options.crosshair_hidden = true; // true|false: hide the crosshair initially?`

`gv_options.map_opacity_control = true; // true|false`

`gv_options.map_type_control = {}; // widget to change the background map`

`gv_options.map_type_control.style = 'menu'; // 'menu'|'none'`

`gv_options.map_type_control.filter = false; // true|false: when map loads, are irrelevant maps ignored?`

`gv_options.map_type_control.excluded = []; // comma-separated list of quoted map IDs that will never show in the list ('included' also works)`

`gv_options.measurement_tools = { visible:false, distance_color:"", area_color:"", position:[] };`

`gv_options.infobox_options = {}; // options for a floating info box (id="gv_infobox"), which can contain anything`

`gv_options.infobox_options.enabled = true; // true|false: enable or disable the info box altogether`

`gv_options.infobox_options.position = ['RIGHT_BOTTOM',180,38]; // [Google anchor name, relative x, relative y]`

`gv_options.infobox_options.draggable = true; // true|false: can it be moved around the screen?`

`gv_options.infobox_options.collapsible = true; // true|false: can it be`

collapsed by double-clicking its top bar?

```
// track-related options:
  // marker-related options:
  gv_options.default_marker =
{ color:'red',icon:'blankcircle',scale:0.1,opacity:1.0 }; // icon can be a URL, but be sure
to also include size:[w,h] and optionally anchor:[x,y]
  gv_options.marker_tooltips = true; // do the names of the markers show
up when you mouse-over them?
  gv_options.marker_shadows = true; // true|false: do the standard
markers have "shadows" behind them?
  gv_options.marker_link_target = '_blank'; // the name of the window or
frame into which markers' URLs will load
  gv_options.info_window_width = 250; // in pixels, the width of the
markers' pop-up info "bubbles" (can be overridden by 'window_width' in individual
markers)
  gv_options.thumbnail_width = 0; // in pixels, the width of the markers'
thumbnails (can be overridden by 'thumbnail_width' in individual markers)

  gv_options.photo_size = [0,0]; // in pixels, the size of the photos in info
windows (can be overridden by 'photo_width' or 'photo_size' in individual markers)

  gv_options.hide_labels = false; // true|false: hide labels when map first
loads?
  gv_options.labels_behind_markers = false; // true|false: are the labels
behind other markers (true) or in front of them (false)?
  gv_options.label_offset = [0,0]; // [x,y]: shift all markers' labels (positive
numbers are right and down)
  gv_options.label_centered = false; // true|false: center labels with
respect to their markers? (label_left is also a valid option.)
  gv_options.driving_directions = false; // put a small "driving directions"
form in each marker's pop-up window? (override with dd:true or dd:false in a
marker's options)
  gv_options.garmin_icon_set = 'gpsmap'; // 'gpsmap' are the small 16x16
icons; change it to '24x24' for larger icons
  gv_options.marker_list_options = {}; // options for a dynamically-created
list of markers
  gv_options.marker_list_options.enabled = true; // true|false: enable or
```

disable the marker list altogether

```
gv_options.marker_list_options.floating = true; // is the list a floating box
```

inside the map itself?

```
gv_options.marker_list_options.position = ['RIGHT_BOTTOM',6,38]; //
```

floating list only: position within map

```
gv_options.marker_list_options.min_width = 160; // minimum width, in
```

pixels, of the floating list

```
gv_options.marker_list_options.max_width = 160; // maximum width
```

```
gv_options.marker_list_options.min_height = 0; // minimum height, in
```

pixels, of the floating list

```
gv_options.marker_list_options.max_height = 300; // maximum height
```

```
gv_options.marker_list_options.draggable = true; // true|false, floating
```

list only: can it be moved around the screen?

```
gv_options.marker_list_options.collapsible = true; // true|false, floating
```

list only: can it be collapsed by double-clicking its top bar?

```
gv_options.marker_list_options.include_tickmarks = false; // true|false:
```

are distance/time tickmarks included in the list?

```
gv_options.marker_list_options.include_trackpoints = false; // true|false:
```

are "trackpoint" markers included in the list?

```
gv_options.marker_list_options.dividers = false; // true|false: will a thin
```

line be drawn between each item in the list?

```
gv_options.marker_list_options.desc = false; // true|false: will the
```

markers' descriptions be shown below their names in the list?

```
gv_options.marker_list_options.icons = true; // true|false: should the
```

markers' icons appear to the left of their names in the list?

```
gv_options.marker_list_options.thumbnails = false; // true|false: should
```

markers' thumbnails be shown in the list?

```
gv_options.marker_list_options.folders_collapsed = false; // true|false: do
```

folders in the list start out in a collapsed state?

```
gv_options.marker_list_options.folders_hidden = false; // true|false: do
```

folders in the list start out in a hidden state?

```
gv_options.marker_list_options.collapsed_folders = [1,2,3]; // an array of
```

folder names or numbers

```
gv_options.marker_list_options.hidden_folders = ['Erythronium
```

```
grandiflorum']; // an array of folder names or numbers
```

`gv_options.marker_list_options.count_folder_items = false; // true|false:`
list the number of items in each folder?

`gv_options.marker_list_options.wrap_names = true; // true|false:` should
marker's names be allowed to wrap onto more than one line?

`gv_options.marker_list_options.unnamed = '[unnamed]'; // what 'name'`
should be assigned to unnamed markers in the list?

`gv_options.marker_list_options.colors = false; // true|false:` should the
names/descs of the points in the list be colorized the same as their markers?

`gv_options.marker_list_options.default_color = ""; // default HTML color`
code for the names/descs in the list

`gv_options.marker_list_options.limit = 0; // how many markers to show in`
the list; 0 for no limit

`gv_options.marker_list_options.center = false; // true|false:` does the
map center upon a marker when you click its name in the list?

`gv_options.marker_list_options.zoom = false; // true|false:` does the map
zoom to a certain level when you click on a marker's name in the list?

`gv_options.marker_list_options.zoom_level = 12; // if 'zoom' is true, what`
level should the map zoom to?

`gv_options.marker_list_options.info_window = true; // true|false:` do info
windows pop up when the markers' names are clicked in the list?

`gv_options.marker_list_options.url_links = false; // true|false:` do the
names in the list become instant links to the markers' URLs?

`gv_options.marker_list_options.toggle = false; // true|false:` does a
marker disappear if you click on its name in the list?

`gv_options.marker_list_options.help_tooltips = false; // true|false:` do
"tooltips" appear on marker names that tell you what happens when you click?

`gv_options.marker_list_options.id = 'gv_marker_list'; // id of a DIV tag`
that holds the list

`gv_options.marker_list_options.header = ""; // HTML code; be sure to put`
backslashes in front of any single quotes, and don't include any line breaks


```

gv_options.marker_list_options.footer = ""; // HTML code

gv_options.marker_filter_options = {}; // options for removing waypoints
that are out of the current view
gv_options.marker_filter_options.enabled = false; // true|false: should
out-of-range markers be removed?
gv_options.marker_filter_options.movement_threshold = 8; // in pixels,
how far the map has to move to trigger filtering
gv_options.marker_filter_options.limit = 0; // maximum number of
markers to display on the map; 0 for no limit
gv_options.marker_filter_options.update_list = true; // true|false: should
the marker list be updated with only the filtered markers?
gv_options.marker_filter_options.sort_list_by_distance = false; //
true|false: should the marker list be sorted by distance from the center of the map?

gv_options.marker_filter_options.min_zoom = 0; // below this zoom
level, don't show any markers at all
gv_options.marker_filter_options.zoom_message = ""; // message to put
in the marker list if the map is below the min_zoom threshold

// gv_options.synthesize_fields = { folder:'<span
class="{folder}">{folder}<'+'/span>' }; // for example: {label:'{name}'} would cause all
markers' names to become visible labels

// Load GPS Visualizer's Google Maps functions (this must be loaded
AFTER gv_options are set):
if (window.location.toString().indexOf('https://') == 0) { // secure pages
require secure scripts

document.write('<scr'+ipt type="text/javascript"
src="https://gpsvisualizer.com/google_maps/functions3.js"></scr'+ipt>');

} else {

document.write('<scr'+ipt type="text/javascript"
src="http://maps.gpsvisualizer.com/google_maps/functions3.js"></scr'+ipt>');

}

```

```

</script>
<style type="text/css">
  /* Put any custom style definitions here
(e.g., .gv_marker_info_window, .gv_marker_info_window_name, .gv_marker_list_ite
m, .gv_tooltip, .gv_label, etc.) */
  #gmap_div .gv_marker_info_window {
    font-size:11px !important;
  }
  #gmap_div .gv_label {
    background-color:transparent; border:0px solid transparent; padding:0px;

    color:black; font:12px Verdana,sans-serif !important; font-
weight:normal !important;
    opacity:1.0; filter:alpha(opacity=100);
  }

</style>
<script type="text/javascript">
  function GV_Map() {

GV_Setup_Map();

  // Track #1 ("Ethiopia---Ethiopia")
  t = 1;trk[t] = {info:[],segments:[]};
  trk[t].info.name = 'Ethiopia---Ethiopia'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
  trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
  trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
  trk[t].segments.push({ points:[ [2,93],[2,93] ] }); // track 1 segment 1
  GV_Draw_Track(t);

  // Track #2 ("Ethiopia---Nigeria")
  t = 1;trk[t] = {info:[],segments:[]};
  trk[t].info.name = 'Ethiopia---Nigeria'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
  trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
  trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color

```

```

= '#CC0000'; trk[t].info.fill_opacity = 0;
  trk[t].segments.push({ points: [ [2,93],[8,38] ] }); // track 1 segment 1
  GV_Draw_Track(t);

// Track #3 ("Nigeria---Nigeria")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Nigeria---Nigeria'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
  trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
  trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
  trk[t].segments.push({ points: [ [8,38],[8,38] ] }); // track 1 segment 1
  GV_Draw_Track(t);

// Track #4 ("India---India")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'India---India'; trk[t].info.desc = 'N    = 3.5'; trk[t].info.clickable
= true; trk[t].info.geodesic = true;
  trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
  trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
  trk[t].segments.push({ points: [ [20,33],[20,33] ] }); // track 1 segment 1
  GV_Draw_Track(t);

// Track #5 ("India---Malaysia")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'India---Malaysia'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
  trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
  trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
  trk[t].segments.push({ points: [ [20,33],[26,93] ] }); // track 1 segment 1
  GV_Draw_Track(t);

// Track #6 ("US---Canada")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'US---Canada'; trk[t].info.desc = 'N    = 3.5'; trk[t].info.clickable
= true; trk[t].info.geodesic = true;

```

```
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [14,100],[32,36] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #7 ("US---India")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'US---India'; trk[t].info.desc = 'N = 3.5'; trk[t].info.clickable =
true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [14,100],[20,33] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #8 ("Canada---Canada")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Canada---Canada'; trk[t].info.desc = 'N = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [32,36],[32,36] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #9 ("Malaysia---Malaysia")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Malaysia---Malaysia'; trk[t].info.desc = 'N = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [26,93],[26,93] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #10 ("China---China")
t = 1;trk[t] = {info:[],segments:[]};
```

```
trk[t].info.name = 'China---China'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [74,94],[74,94] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #11 ("China---US")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'China---US'; trk[t].info.desc = 'N    = 3.5'; trk[t].info.clickable =
true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [74,94],[14,100] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #12 ("Thailand---Thailand")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Thailand---Thailand'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [68,33],[68,33] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #13 ("Portugal---Portugal")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Portugal---Portugal'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [62,91],[62,91] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #14 ("South_Korea---South_Korea")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'South_Korea---South_Korea'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [50,93],[50,93] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #15 ("US---US")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'US---US'; trk[t].info.desc = 'N    = 3.5'; trk[t].info.clickable =
true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [14,100],[14,100] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #16 ("Spain---Spain")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Spain---Spain'; trk[t].info.desc = 'N    = 3.5'; trk[t].info.clickable
= true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [44,32],[44,32] ] }); // track 1 segment 1
GV_Draw_Track(t);
```

```
// Track #17 ("Taiwan---Taiwan")
t = 1;trk[t] = {info:[],segments:[]};
trk[t].info.name = 'Taiwan---Taiwan'; trk[t].info.desc = 'N    = 3.5';
trk[t].info.clickable = true; trk[t].info.geodesic = true;
trk[t].info.color = '#f7d2f5'; trk[t].info.width = 4; trk[t].info.opacity = 0.9;
trk[t].info.outline_color = '#f7d2f5'; trk[t].info.outline_width = 0; trk[t].info.fill_color
= '#CC0000'; trk[t].info.fill_opacity = 0;
trk[t].segments.push({ points:[ [38,94],[38,94] ] }); // track 1 segment 1
```

```
GV_Draw_Track(t);
```

```
GV_Draw_Marker({lat:74,lon:94,name:'China',desc:'(1,100)(74,94)<ul
class="more_info"><li> <a target="_blank"
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font color=red>China</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:68,lon:33,name:'Thailand',desc:'(1,100)(68,33)<ul
class="more_info"><li> <a target="_blank"
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>Thailand</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:62,lon:91,name:'Portugal',desc:'(1,100)(62,91)<ul
class="more_info"><li> <a target="_blank"
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>Portugal</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:56,lon:35,name:'Germany',desc:'(1,100)(56,35)<ul
class="more_info"><li> <a target="_blank"
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>Germany</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:50,lon:93,name:'South_Korea',desc:'(1,100)(50,93)<ul
class="more_info"><li> <a target="_blank"
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>South_Korea</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:44,lon:32,name:'Spain',desc:'(1,100)(44,32)<ul
class="more_info"><li> <a target="_blank"
```

```
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font color=#000000>Spain</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:38,lon:94,name:'Taiwan',desc:'(1,100)(38,94)<ul
class="more_info"><li><a target="_blank"
href=></a></li></ul>',color:'#96ab03',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>Taiwan</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:32,lon:36,name:'Canada',desc:'(2,100)(32,36)<ul
class="more_info"><li><a target="_blank"
href=></a></li></ul>',color:'#6495ED',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font color=red>Canada</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:26,lon:93,name:'Malaysia',desc:'(2,100)(26,93)<ul
class="more_info"><li><a target="_blank"
href=></a></li></ul>',color:'#6495ED',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>Malaysia</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:20,lon:33,name:'India',desc:'(2,100)(20,33)<ul
class="more_info"><li><a target="_blank"
href=></a></li></ul>',color:'#6495ED',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font color=#000000>India</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:14,lon:100,name:'US',desc:'(2,100)(14,100)<ul
class="more_info"><li><a target="_blank"
href=></a></li></ul>',color:'#6495ED',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font color=#000000>US</font>',folder:'methods'}});
```

```
GV_Draw_Marker({lat:8,lon:38,name:'Nigeria',desc:'(7,100)(8,38)<ul
class="more_info"><li><a target="_blank"
```



```
href=></a></li></ul>',color:'#FFBF00',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font color=red>Nigeria</font>',folder:'methods'});
```

```
GV_Draw_Marker({lat:2,lon:93,name:'Ethiopia',desc:'(7,100)(2,93)<ul
class="more_info"> <li> <a target="_blank"
href=></a></li></ul>',color:'#FFBF00',icon:'',scale:0.6,url:'',label:'<div
id=blueDIV><font size=3><b><font
color=#000000>Ethiopia</font>',folder:'methods'});
```

```
GV_Build_And_Place_Draggable_Box({base_id:'gv_infobox2',class_name:'gv_infobox
',position:['TOP_LEFT',74,4],draggable:true,collapsible:true});
}
GV_Map(); // execute the above code
</script>
</body>
</html>
```

36.Wordcloud
43.Wordcloud2
44.Spider plot
45.Plot for Two Way Anova
5. Detatitica

(Clear/Type/Submit to see data format)

Clear

```
Source,Target,WCD
Canada,U.S.,2
China,Portugal,1
China,South_Korea,1
China,Thailand,4
Ethiopia,Nigeria,1
China,Germany,1
India,U.S.,1
India,Malaysia,1
China,Spain,1
China,Taiwan,1
China,U.S.,1
Portugal,U.S.,1
```

lots in R

Chinese notes

Submit for Wordcloud2

60A4.*CircleBar. to #15 CircleBar

60A5.*CirclePacking. to #34 CirclePacking

60A6.D4.*HeatDendro. to #52 HeatDendr

60A7.D55.*Network. next to #46

60A4.^CircleBar. to #15 CircleBar

60A5.*CirclePacking. to #34 CirclePacking.

Submit

Tips for R The number of Clusers is set at (2,targe, value) = 2, for example

Copy & Paste the Code in R to Rstudio

Select text

```
individual,value
China,10
Thailand,4
Portugal,2
Germany,1
South_Korea,1
Spain,1
Taiwan,1
Canada,2
Malaysia,1
India,2
U.S.,5
Nigeria,1
Ethiopia,1
```

Thailand,4
1.Refresh
2.Paste to input box
3.Click
ChatGPT Writer
Next to:Googlemaps
(Clear/Type/Submit to see data format)

Clear

```
individual,value  
China,10  
Thailand,4  
Portugal,2  
Germany,1  
South_Korea,1  
Spain,1  
Taiwan,1  
Canada,2  
Malaysia,1  
India,2  
U.S,5  
Nigeria,1  
Ethiopia,1
```

to draw plots in R

Chinese notes

Submit for Authorcorrence5

COA7B1 D1:4 to Network for SNA

6/11/21

33.Multi-Line Bar2 Chart

36.Wordcloud

43.Wordcloud2

44.Spider plot

45.Plot for Two Way Anova

F.Rstatistics

33.Multi-Line Bar2 Chart

36.Wordcloud

Submit

Tips for R

Copy & Paste the Code in R to Rstudio

Select text

```
library(wordcloud2)
library(tm)
library(wordcloud)
library(SnowballC)
data <- data.frame(individual=c("China","Thailand","Portugal","Germany","South_Korea","Spain","Taiwan","Canada","Malaysia","India","U.S","Nigeria","Ethiopia"),
,value=c(10,4,2,1,1,1,1,1,2,1,2,5,1,1))
)
wordcloud2(data, size = 1, minSize = 0, gridSize = 0, fontFamily = 'Segoe UI', fontWeight = 'bold', color = 'random-dark', backgroundColor = "white", minRotation = -pi/4, maxRotation = pi/4, shuffle = TRUE, rotateRatio = 0.4, shape = 'circle', ellipticity = 0.65, widgetsize = NULL, figPath = NULL, hoverFunction = NULL)
#wordcloud2(demoFreq, size=1.6, color='random-light', backgroundColor="black")
```



```

library(highcharter)
library(dplyr)
data <-
data.frame(individual=c("China","Thailand","Portugal","Germany","South_Korea","S
pain","Taiwan","Canada","Malaysia","India","U.S","Nigeria","Ethiopia")
,value=c(10,4,2,1,1,1,1,2,1,2,5,1,1)
)

highchart() %>%
  hc_title(text = "World Cloud") %>%
  hc_add_series(data = data,type = "wordcloud",name= "Score",hcaes(name
=data$individual,weight = data$value)) %>%
  hc_add_theme(hc_theme_flat())

```



```

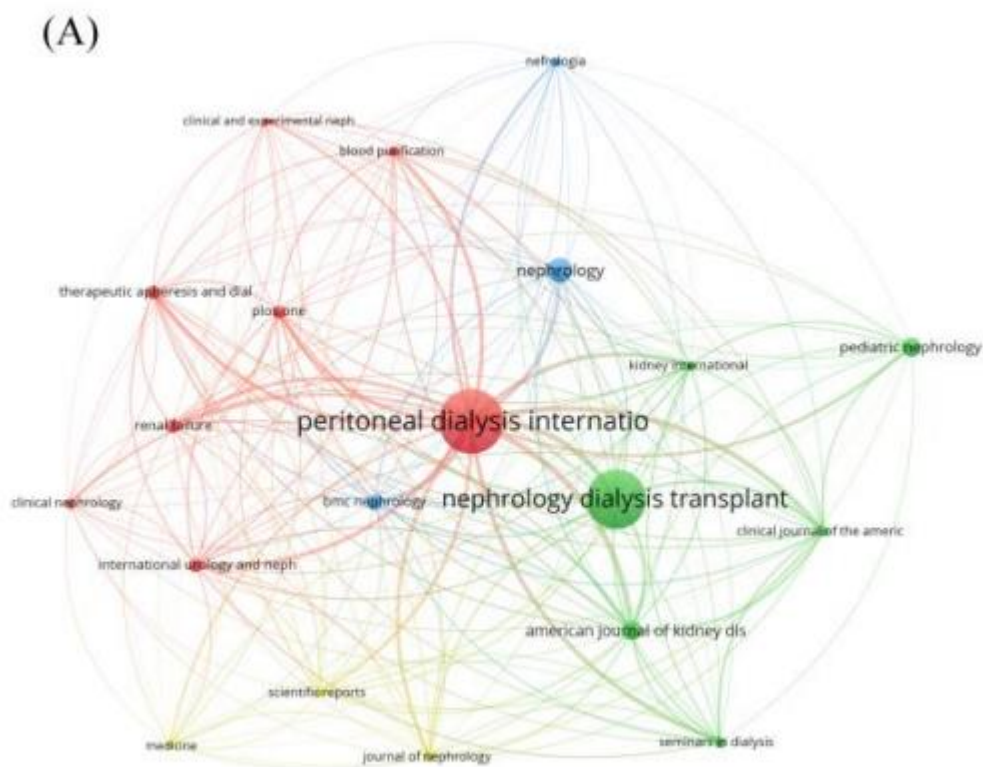
library(wordcloud2)
library(tm)
library(wordcloud)
library(SnowballC)
data <-
data.frame(individual=c("China","Thailand","Portugal","Germany","South_Korea","S
pain","Taiwan","Canada","Malaysia","India","U.S","Nigeria","Ethiopia")

```

,value=c(10,4,2,1,1,1,1,2,1,2,5,1,1)

)

```
wordcloud2(data, size = 1, minSize = 0, gridSize = 0, fontFamily = 'Segoe UI',  
fontWeight = 'bold', color = 'random-dark', backgroundColor = "white", minRotation  
= -pi/4, maxRotation = pi/4, shuffle = TRUE, rotateRatio = 0.4, shape = 'circle',  
ellipticity = 0.65, widgetsize = NULL, figPath = NULL, hoverFunction = NULL)  
wordcloud2(data, size=0.6, color='random-light', backgroundColor="black")
```



Top 20 journals collaboration relationship

Analysis of journals and co-cited journals

The top 10 journals and co-cited journals that published the highest number of articles in the direction of XXXX were shown in Table 2. The top 20 journals and their partnerships were mapped using VOSviewer (Figure 2A). The process of constructing the relationship between journals by VOSviewer is mainly based on the analysis of co-words in the literature data, such as common keywords, common research themes, and so on. Firstly, VOSviewer collects the data in the literature and

transforms them into the co-word matrix, then, the similarity calculation is carried out based on the co-word matrix, and after that, the visualization display is completed by the clustering algorithm and the layout algorithms such as multi-dimensional scaling. Next, themes were assigned to articles. In groups of themes, the journal occurrences were then observed and analyzed by social network and cluster analyses, as shown in panel A of Figure 2.

As can be seen from the figure, most of In review the top ten active journals are specialized journals in the fields of XXXX disease and XXXX. XXXX International was the most active journals in this field. It's followed by Nephrology Dialysis Transplantation and Nephrology.

Figure 2B is a network of co-cited journals showing the relationships between the co-cited journals involved in the references. Peritoneal Dialysis International (4420) is the most cited journal. This is followed by Nephrology Dialysis Transplantation (4037) and Kidney International (3791). Among the top ten co-cited journals, New England Journal of Medicine ranks 9th with 1294 articles. High-scoring journals are considered important journals with high co-citations. The journal with the highest centrality is Journal of the American Society of Nephrology (0.02).

If the authors' explanation for Panel A in Figure 2 is as follows, we can acknowledge the logical rationale provided:

1. A co-word analysis was conducted using both keywords and words from titles across all articles.
2. Primary themes were identified based on the co-word analysis from step 1.
3. Each article was then categorized according to the themes identified in step 2, as supported by a previous study available at <https://pubmed.ncbi.nlm.nih.gov/36930064/>.
4. Based on the collection of journals associated with articles of particular themes, multiple journals were grouped under their corresponding themes. The frequency of journal appearances within these thematic groups was noted.
5. Social network analysis, in conjunction with cluster analysis, was employed to examine the interrelations of themes among journals.

Crucially, this process maps the relationships of journals based on article themes, rather than illustrating collaboration relationships between journals.

<https://pubmed.ncbi.nlm.nih.gov/?term=Brin+Mitchell+F.%5BAuthor%5D&sort=date>

Save Email Send to Sort by: Most recent

Save citations to file

Selection:

Format:

← ChatGPT Writer



New Analysis

Search PubMed

Search by author name, title, MeSH keyword, etc

OR

Upload a spreadsheet of PMIDs

未選擇任何檔案

OR

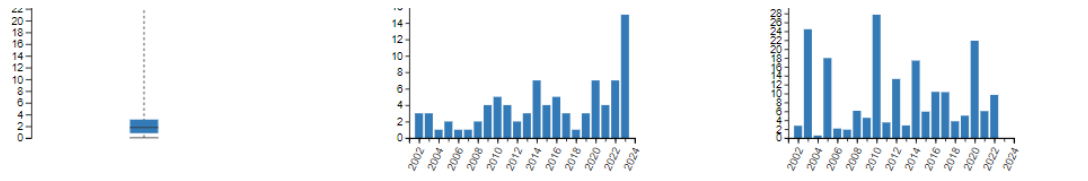
Input a list of PMIDs

13916521
16299091
17496074

A maximum of 10000 PMIDs may be queried at a time.

The *iCite* database currently contains articles published from 1980 to present.

https://icite.od.nih.gov/analysis#



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ou make changes below, the summary information and charts above are updated.

1 2002 To Only research articles Only papers cited by clinical articles Only clinical articles Clear Filters Total Pubs: 87 Export

PMID	Year	Title	Authors	Journal	NIH Percentile	RCR
12112...	2002	Severe generalized dystonia due to primary putaminal degeneration: case repor	Ruth H Walker, Dushyant P Pur	Mov Disord	4.9	0.10
12195...	2002	Botulinum toxin type A (BOTOX) for treatment of migraine.	William J Binder, Mitchell F Brir	Dis Mon	52.0	1.05
12402...	2002	Epsilon-sarcoglycan mutations found in combination with other dystonia gene r	Christine Klein, Liu Liu, Dana D	Ann Neurol	67.2	1.59
12756...	2003	A heteroplasmic mitochondrial complex I gene mutation in adult-onset dystonia	David K Simon, Jennifer Friedm	Neurogenetics	32.1	0.59

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	PMID	Year	Title	Authors	Journal	Article	RCR	Provisiona	Citations P	Expected C	Field Citat	NIH Perce	Human
2	12112210	2002	Severe gen	Ruth H We	Mov Disor	No	0.1	No	0.190476	1.840263	4.291383	4.9	1
3	12195263	2002	Botulinum	William J	Dis Mon	No	1.05	No	1.714286	1.633292	3.439135	52.0	1
4	12402271	2002	Epsilon-sa	Christine K	Ann Neurol	Yes	1.59	No	3.428571	2.155818	5.590759	67.2	0.5
5	12756609	2003	A heteropl	David K S	Neurogene	Yes	0.59	No	1.45	2.444793	6.538145	32.1	0.57
6	12834116	2003	Botulinum	Lucian Sul	Ann Otol F	Yes	0.75	No	1	1.33184	2.130321	39.8	1
7	12953276	2003	A double-t	C Warren	Ann Neurol	Yes	23.1	No	52.65	2.279431	5.883233	99.6	0.67
8	15022181	2004	Corticospir	Colum D M	Mov Disor	Yes	0.54	No	1	1.841732	3.890305	29.4	1
9	15836567	2005	Botulinum	David W L	Headache	Yes	4.98	No	8.222222	1.650548	3.119544	93.1	1
10	15947626	2005	Botulinum	Brigitte Sc	J Urol	Yes	12.98	No	21	1.618025	2.997506	98.7	1
11	16731213	2006	Pooled ana	Catherine C	Arch Phys	Yes	2.11	No	3.176471	1.503531	2.52062	76.3	1
12	17617291	2007	Formation	Stuart A Y	Clin Ther	Yes	1.86	No	3.125	1.683519	2.936383	72.3	1
13	18175340	2008	Phenotypic	Deborah R	Mov Disor	Yes	1.03	No	2.333333	2.275604	4.964933	51.2	1
14	18546321	2008	Long-term	Mitchell F	Mov Disor	Yes	5.07	No	8.666667	1.710607	3.149483	93.3	0.71
15	19341758	2009	Basic and c	Mitchell F	Toxicol	Yes	0.36	No	0.785714	2.15353	4.501903	19.4	1
16	19470342	2009	Developme	Mitchell F	Toxicol	No	0.81	No	1.5	1.853443	3.573788	42.5	1
17	19626521	2009	Difference	Mitchell F	J Cosmet L	No	0.03	No	0.071429	2.048561	4.177253	2.0	1
18	19744746	2009	Safety and	Mitchell F	J Am Acad	Yes	3.32	No	5.5	1.656637	2.965104	87.1	1
19	20487038	2010	Onabotulin	David W L	Headache	Yes	21.57	No	42.53846	1.972175	3.687354	99.5	1
20	20669276	2010	Genetic ev	Nutan Sha	Mov Disor	Yes	0.67	No	1.615385	2.425609	5.158278	35.8	1
21	20737546	2010	Meta-analy	Markus Na	Mov Disor	Yes	4.33	No	7.307692	1.689153	2.769242	91.4	1

	A	B	C	D	E	F	G	H	I	J
1	C:\Users\user\ID	4318	Total Citati	Year	PMID	Cited	By	PMIDs		
2	savedreccs (70)	39	4	2002	12112210	16409555	12869809	22136648	14534912	
3		56	36	2002	12195263	192	Takeout DOI	27	cited authors	24841626
4		22	72	2002	12402271	153		95		19535332
5		23	29	2003	12756609	26306407	16949108	30055843	31452146	32470904
6		37	20	2003	12834116	204	Journal only	22669	category	5062687
7		37	1053	2003	12953276	273		15630		5739547
8		23	19	2004	15022181	16532448	19339733	16731028	20077474	17936390
9		49	148	2005	15836567	28884716	17194546	16618279	16820222	19835236
10		88	378	2005	15947626	19361151	24100600	21509809	34977952	17313422
11		42	54	2006	16731213	261	Remove not Self	31078139	28118416	20178412
12		45	50	2007	17617291	18784398	20467038	31811879	34558213	24379687
13		51	35	2008	18175340	2262	Remove space	35	To links journ	23401150
14		41	130	2008	18546321	3015		25		799
15		38	11	2009	19341758	36287978	20921324	33837264	28322922	32884513
16		39	21	2009	19470342	20811177	21353660	28380276	26114235	30217070
17		99	1	2009	19626521	3154	Sorting	2888	Coupling each	090
18		14	77	2009	19744746	3169				
19		38	553	2010	20487038	29404713	30869961	2613		9136
20		39	21	2010	20669276	220	To links	22773857	27188707	30643666
21		56	95	2010	20737546	368			25414799	24379687

< > ... journal | data | doicited | cluster | links | pai ... + : 啟用 Winc

2009	27.5	4	110	2	2
2010	138.5	5	690	2	2
2011	17.5	4	70	2	2
2012	130.5	2	281	2	2
2013	18.33	3	55	2	2
2014	46.57	7	326	2	2
2015	22	4	88	2	2
2016	32.65	5	163	2	2
2017	48.67	3	148	2	2
2018	38	1	38	2	2
2019	16	3	48	2	2
2020	20.57	7	144	2	2
2021	7.25	4	29	2	2
2022	4.71	7	33	2	2
2023	0	15	0	2	2

Ranked by Pubs by year in columns: Weighted RCR by Year, Median RCR by Year, Total Citations by Publication Year, and Citations per Pub by Year with blanks fro

3488784	2022	7	20	0.33	29
34832787	2022	8	23	0.33	1
36751871	2023	0	0	0.33	1
36849797	2023	0	17	0.33	3
37137724	2023	0	24	0.33	30
37235376	2023	0	33	0.33	31
37499076	2023	0	37	0.33	32
37499077	2023	0	44	0.32	33
37499078	2023	0	51	0.32	34
37499079	2023	0	52	0.32	35
37499080	2023	0	59	0.32	36
37499081	2023	0	61	0.32	37
37499082	2023	0	63	0.32	38
37499083	2023	0	67	0.32	39
37499085	2023	0	68	0.32	40
37499087	2023	0	70	0.32	41
37499089	2023	0	72	0.32	42

PMID, Year, RCR, NIH Percentile, Total Citations, AWS, SCI_JIF

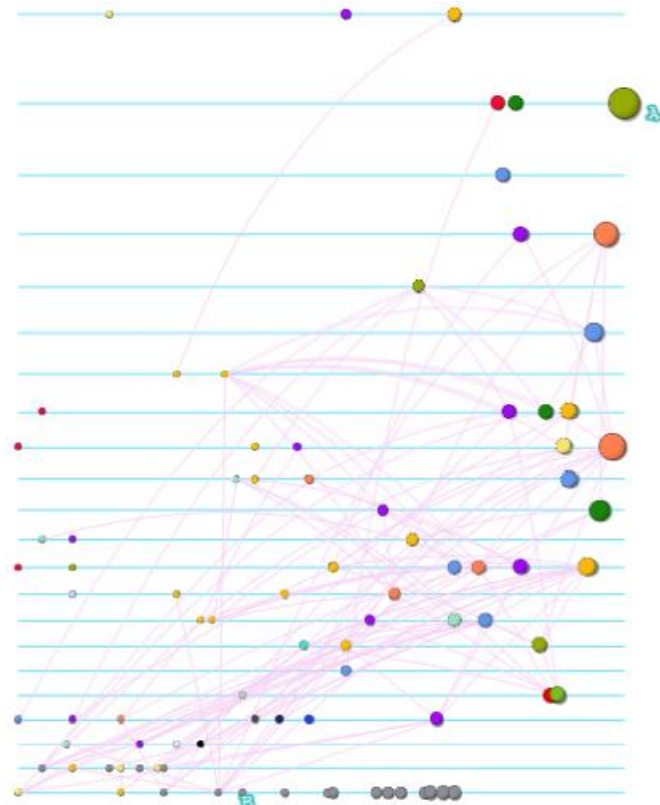
28387038	34932787	1
28387038	31102144	1
28389378	35546268	1
28389376	34448754	1
28389376	32602955	1
28389376	36751871	1
27910297	31102144	1
29219190	29219190	0
31494402	31494402	0
31102144	31102144	0
31475573	35546268	1
31475573	33200944	1
31475573	34932787	1
31475573	34448754	1
33086749	33086749	0
32170665	32170665	0
32170665	32170665	0

All-citation= 4227 Self-citation 123 Article with most citation= 37235376

x-index on column [Grouping in last column]

Submit

2002	37	3	112
2003	367	3	1102
2004	19	1	19
2005	263	2	526
2006	54	1	54
2007	50	1	50
2008	82	2	165
2009	28	4	110
2010	138	5	690
2011	18	4	70
2012	130	2	261
2013	18	3	55
2014	47	7	326
2015	22	4	88
2016	33	5	163
2017	49	3	146
2018	36	1	36
2019	16	3	48
2020	21	7	144
2021	7	4	29
2022	5	7	33
2023	0	15	0



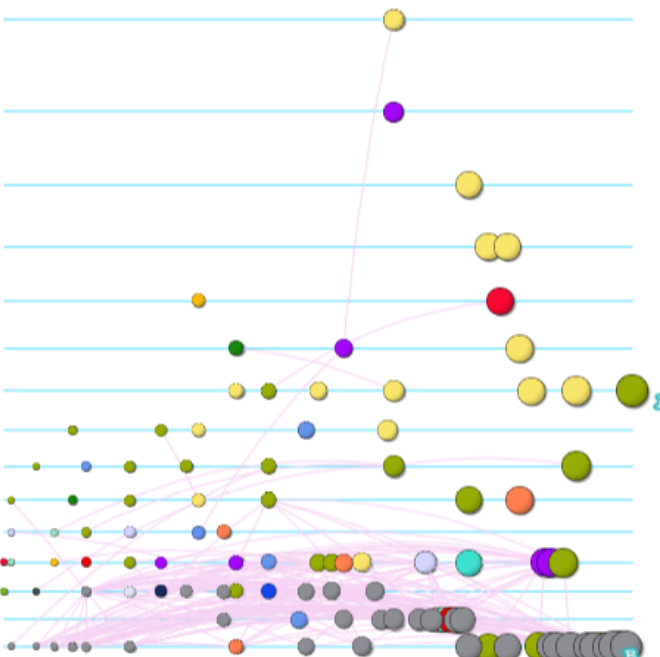
<https://pubmed.ncbi.nlm.nih.gov/?term=Brin+Mitchell+F.%5BAuthor%5D&sort=date>

Brin Mitchell F.[Author]

<https://pubmed.ncbi.nlm.nih.gov/?term=Willy+Chou%5BAuthor%5D&sort=date>

Willy Chou[Author]

2007	30	1	30
2008	27	1	27
2009	25	1	25
2011	38	2	75
2013	13	2	26
2014	12	3	36
2015	12	7	84
2016	7	5	36
2017	9	7	63
2018	8	7	54
2019	6	6	34
2020	12	17	203
2021	9	15	139
2022	5	12	58
2023	1	23	19



<https://www.healthup.org.tw/raschonline/medicine2023.htm>

1.Refresh

2.Paste to input box

3.Click on Me

Next to:Network7A12

(Clear/Type/Submit to see data format)

Clear

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draw plots in R

Chinese notes

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 - 2.Paste to input box
 - 3.Click on Me
- Next to:Network7A12

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Clear

```
Leader, follower, cluster, WCD, clusterB2
China, Thailand, 1, 4, 1
Canada, U.S, 2, 2, 2
China, South_Korea, 1, 1, 1
China, Portugal, 1, 1, 1
Ethiopia, Nigeria, 3, 1, 3
China, Germany, 1, 1, 1
India, U.S, 2, 1, 2
India, Malaysia, 2, 1, 2
China, Spain, 1, 1, 1
China, Taiwan, 1, 1, 1
China, U.S, 1, 1, 2
Portugal, U.S, 1, 1, 2
China, 10, 1
U.S, 5, 2
```

pts in R

Chinese notes

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Two times on clicking bottoms 1,2,3

60AA3.D5:nodes & relations Group in Net.

Submit

Tips for R

Copy & Paste the Code in R to Rstudio

Select text

```
value2=c(2,5,10,1,1,2,1,1,1,1,2,4))
#nodes$value2 used for computing original counts in textabc2 set
# data<-as.matrix.data.frame(data)
# df.g <- graph.data.frame(d = data[,1:2], directed = FALSE)
# plot(df.g, vertex.label = V(df.g)$name,vertex.color = "green")
# Turn it into igraph object
network <- graph_from_data_frame(d=data, vertices=nodes, directed=F)
# Make a palette of 3 colors
library(RColorBrewer)
coul <- brewer.pal(3, "Set1")
# Create a vector of color
my_color <- coul[as.numeric(as.factor(V(network)$carac))]
deg <- degree(network, mode="all")
V(network)$size <- nodes$value2*3 #deg*3
# We could also use the audience size value:
#V(network)$size <- nodes$value*1.0
# nodes are sized by value of WCD
```

啟用 Window

Copy R-code to R for visualizations

```
plot(network)$value*1/20 and V(network)$size <- deg*3 for edge a
ble sizes
end(x=-1.1, y=-1.1, c("Newspaper", "Television", "Online News"), pch
="#777777", pt.bg=my_color, pt.cex=2.5, bty="n", ncol=1)
ps://oak.ucc.nau.edu//yc424/_book/data-to-network.html
t(network, vertex.shape="none", vertex.label=V(net)$media,
vertex.label.font=2, vertex.label.color="gray40",
vertex.label.cex=.7, edge.color="gray85")

e.start <- ends(network, es=E(network), names=F)[,1]
e.col <- V(network)$color[edge.start]

t(network, edge.color=edge.col, edge.curved=.1)
<- cluster_edges_betweenness(network)
```

