Exploring Biological Networks in 3D, Stereoscopic 3D, and Immersive 3D with iCA VE

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Biological networks are becoming increasingly large and complex, pushing the limits of existing 2D tools. iCA VE is an open-source software tool for interactive visual explorations of large and complex networks in 3D, stereoscopic 3D, or immersive 3D. It introduces new 3D network layout algorithms and 3D extensions of popular 2D network layout, clustering, and edge bundling algorithms to assist researchers in understanding the underlying patterns in large, multi-layered, clustered, or complex networks. This protocol aims to guide new users on the basic functions of iCA VE for loading data, laying out networks (single or multi-layered), bundling edges, clustering networks, visualizing clusters, visualizing data attributes, and saving output images or videos. It also provides examples on visualizing networks constrained in physical 3D space (e.g., proteins; neurons; brain). It is accompanied by a new version of iCA VE with an enhanced user interface and highlights new features useful for existing users. © 2018 by John Wiley & Sons, Inc.

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INTRODUCTION

Networks are frequently used to visually communicate and understand complex associations in biomedical data (Merico, Gfeller, & Bader, 2009; M. Newman, 2010). There are a myriad of tools that assist in interactively exploring networks in two dimensions (2D) including Cytoscape (Shannon et al., 2003), Gephi (Bastian, Heymann, & Jacomy, 2009), and Osprey (Breitkreutz, Stark, & Tyers, 2003). However, recent developments in both computational and experimental technologies are leading to increasingly large and heterogeneous data with ever more complex interactions, pushing the boundaries of existing 2D tools, which are inherently limited by the number of interactions that can fit within a 2D screen. New visualization approaches are needed to address the challenges in exploring such datasets (Pavlopoulos et al., 2015). To address these needs, we have recently introduced iCAVE, an open-source software tool for visualization of complex, large, and/or multi-layered networks in 3D, stereoscopic 3D, and immersive 3D (Liluashvili et al., 2017). For networks that represent abstract data, the addition of the third dimension allows for greater flexibility, especially when combined with rotation, zoom, and navigation capabilities (Sollenberger & Milgram, 1993). For networks where node
coordinates have physical meaning within the 3D space (e.g., protein residue-residue or brain connectivity networks), users can navigate within the actual physical layout of their networks.

This protocol aims to guide new users in gaining hands-on experience using the basic functions of iCAVE to interactively explore networks in 3D. Users that follow the protocols will learn how to load data, lay out networks (single or multi-layered), bundle edges, cluster networks, visualize clusters, visualize different data attributes (e.g., colors, sizes, directions), and save iCAVE outputs of network statistics reports, image snapshots, and animation videos. The protocols are illustrated with multiple examples and 2D snapshots of their output images in the accompanying figures. Links for input files are provided in Internet Resources. For a couple of networks, we also provide animated videos. These networks can then be easily shared with colleagues or reproduced.

Briefly, Basic Protocol 1 describes how to customize, load, and navigate network data and to run different network layout algorithms within iCAVE. Basic Protocol 2 details how to visualize and explore interconnections in multi-layered networks and how to bundle edges in order to reveal connectivity strengths between different data types. Basic Protocol 3 provides information on running different network-topology-based graph clustering algorithms, cluster visualization options, and additional applications of edge bundling to reveal intra- and inter-cluster connectivity strengths. Finally, Basic Protocol 4 details how to visualize networks with known 3D coordinates. In addition, Support Protocol 1 provides instructions on installing iCAVE, and Support Protocol 2 details both the steps to turn the stereo option on or off for stereoscopic 3D viewing, and the necessary hardware and software resources. Finally, this manuscript is further accompanied with a new version of iCAVE that has an enhanced user interface, and its new features are summarized in the Background Information section. Overall, this protocol will assist researchers in a variety of fields to visually explore complex networks in 3D and to identify and communicate special structures within them using iCAVE.

**VISUALIZING NETWORK DATA FROM INPUT FILE IN iCAVE**

Users can download the latest version of iCAVE from [http://research.mssm.edu/gumuslab/software.html](http://research.mssm.edu/gumuslab/software.html) and follow the installation instructions provided in Support Protocol 1. After successful installation, all examples provided in this protocol can be generated in 3D using iCAVE. For additional stereo 3D viewing, necessary resources and details on turning stereo on or off are provided in Support Protocol 2.

In Basic Protocol 1, we demonstrate how to customize, load, lay out, navigate, and save statistics and video animations of a network from Goh et al. (2007) on disease-disease and disease-gene associations in a diseasome network (see diseasome.txt in Internet Resources). We use the most common iCAVE input, which is a tab-delimited text file of all edges that is customized to user specifications, although it is also possible to input only a list of gene names for a quick, initial query of iCAVE’s COMBO database (see Liliuashvili et al., 2017, for details of this). The input data are passed as an argument to iCAVE during the execution time, based on which it generates an interactive visualization of the network in 3D space. Users can select from any of its layout algorithms, which include both novel 3D algorithms and extensions of popular 2D algorithms to 3D. The network layout step is important, as it can affect understanding of its structure. Users are encouraged to test all layout algorithms within iCAVE to find the one most suitable for their networks; this will be different for each network. They can then navigate within the network by interactively zooming and rotating, viewing network statistics, or save network animations at user-defined speed, zoom, or viewpoint to communicate the network with colleagues.
**Necessary Resources**

**Hardware and Software**

PC or Mac computer with iCAVE. See Support Protocol 1 for details.

**Input File Format**

In general, iCAVE input is a tab-separated text file with two sections: an optional section (at the top of the file) and a required section (below). The optional section lists the nodes and the required section lists the node pairs that make up the edges.

The optional section begins with the header:

```plaintext
#nodes name
```

The required section starts immediately below (no spaces) with:

```plaintext
#edges node1 node2
```

Both sections may also include user-defined attributes for nodes and edges. iCAVE provides ample node and edge customization options for better network visualizations, explorations and communications in 3D. We provide a complete list of keywords and their descriptions for specifying node attributes in Table 8.27.1 and for specifying edge attributes in Table 8.27.2. Furthermore, iCAVE allows great flexibility in specifying edge and node colors, which are proven quite important in highlighting important parts of a network. A wide range of colors can be specified based on standards available in R software (Ihaka & Gentleman, 1996), which we summarize in `$iCAVE_HOME/vrnet/ColorChart.pdf`.

The input file `diseasome.txt` includes 1,419 nodes and 2,738 edges, and the optional nodes section begins with:

```plaintext
#nodes name label node_scale color_code
55  Deafness 5 mediumorchid1
47  Leukemia 4.75828 mediumorchid1
```

Here, the header line includes the required `name` attribute and the optional `label`, `node_scale`, and `color_code` attributes. Each node has a unique name. Additional user-defined labels are specified under the optional `label` attribute for all or a subset of nodes. The optional `node_scale` attribute enables node size scaling based on user-defined criteria. Finally, the optional `color_code` attribute enables user-defined color specifications for every node. In this example, the two lines below the header line specify nodes that represent diseases, and the last two lines specify nodes that represent genes. Disease nodes have descriptive labels which we will interactively turn on or off during this session, while labels for gene nodes are left blank. Diseases are ranked in decreasing order of their user-defined `node_scale` values (starting from 5) that specify the disease node size based on their scale, while all gene nodes are assigned a `node_scale` value of 1. Here, to differentiate between node types, all disease nodes are user-defined in color `mediumorchid1`, and all gene nodes in `dodgerblue`. 
Table 8.27.1 Specification and Customization of Node Attributes

| Keyword         | Description                                                                                                                                 |
|-----------------|---------------------------------------------------------------------------------------------------------------------------------------------|
| name (required) | Provide the unique name/id to all nodes listed                                                                                               |
| label           | You can specify node labels different than the node name to be displayed. You can also specify labels for a subset of nodes (leave label column blank for nodes without labels). |
| xyz             | $x, y, z$ coordinates of node; can be real or integer values                                                                                   |
| node_color      | To specify a single, unified color for all the nodes. In the next tab, provide the name of the color.                                           |
| color_code      | To specify different colors for various nodes. Provide the name of color for each individual node.                                            |
| texture_type    | To specify texture, provide an integer value accordingly with the following: 0: dashed node, 1: polka dot, 2: triangles in the circle, 3: striped, 4: stars |
| cluster         | Specify an integer id to associate each node with a pre-described cluster.                                                                     |
| foldChange      | Specify this when you want to input gene expression values, where the node color reflects the increase or decrease in expression of the gene the node represents |
| node_scale      | To change the scale of the nodes relative to each other; you can specify any real number value                                               |
| labels_only_    | When used in combination with node_scale, only the labels of scaled nodes (the ones with a different value than 0) are displayed              |
| scaled_nodes    |                                                                                                                                               |
| layer_code      | To specify a multi-layered network, specify the layer (use integer value) of each node                                                   |
| category        | To differentiate between different classes of nodes. If using superpathway, you can use the following codes: 0: protein, 1: complex, 2: rna, 3: mirna, 4: family, 5: abstract. If visualizing metabolic network, you can use the following codes: 0: unclassified, 1: amino acid, 2: acyl_carnitines, 3: diacyl-PC, 4: lyso-PC, 5: acyl-alkyl-PC, 6: sphingomyelin. |
| showNodeProperty| If no node_scale option is provided and you specify this, each node is highlighted at intensity in accordance with its number of neighbors     |

In the diseasome network, the first few lines of the required edges section are:

```
#edges  node1  node2  edge_color  gray96
1285    3211   
468     2914   
416     3825   
```

The header includes the required node1 and node2 attributes and an optional edge_color attribute that assigns the same color to all edges (alternatively, using the color_code option, users can define color for each edge separately). Each line below the header corresponds to an individual edge. In this example, the first edge is specified between nodes 1285 and 3211, and the color for all edges is specified as gray96.
### Table 8.27.2 Specification and Customization of Edge Attributes

| Keyword           | Description                                                                 |
|-------------------|-----------------------------------------------------------------------------|
| node1, node2      | Provide the name/id of nodes that are connected                            |
| isDirected        | Specify this if the network is directed                                    |
| edge_color        | To specify a single, unified color for all the edges. In the next tab, provide the name of the color. |
| color_code        | To specify different colors for various edges. Provide the name of color for each individual edge. |
| weight            | Weight values can be positive or negative real numbers. Negative values are translated into the visualization as the color of the edge being red/white stripes. Positive values are translated into the visualization as the color of the edge being blue/white stripes. The absolute value of the weight is translated as the frequency of the stripes. |
| hideEdgeWeight    | Specify this to hide edge weights in visualization                         |
| thickness         | To scale the thickness of each edge, specify a positive real number        |
| strength          | Specify this to differentiate between edges when visualizing GWAS studies. You can use the following codes: -1: default, 0: if two nodes interact, 1: if they share a disease, 2: if they interact and share a disease. |
| interactionType   | To specify the type of interaction between two nodes. You can use the following codes: -a>: activation, -a|: inhibition, -t>: transcription activation, -t|: transcription inhibition. |
| showNodeProperty  | If you do not provide node information but still want to see node property (as explained in Table 8.27.1) |

### Starting iCAVE

1. Copy the input file diseasome.txt into the $iCAVE_HOME/vrnet/importFiles folder.

2. Open a terminal and set your directory to $iCAVE_HOME/vrnet.

3. Launch iCAVE with the following syntax:

   ```
   vrnetview -Gdiseasome.txt -O10 -l1 -B1
   
   -Gdiseasome.txt to open the graph specified in file diseasome.txt
   -O10 to specify that the file is in iCAVE format.
   -l1 to use weighted-force directed algorithm for network layout. Other layout options are:
   -l0 = Force-directed
   -l2 = Hemispherical
   -l3 = Semantic Levels
   -l4 = LinLog
   -l5 = Weighted LinLog
   -l6 = Hybrid Force-directed (default layout if no layout option is provided)
   -l7 = Simulated Annealing Force-directed
   -l8 = Coarsened Force-directed
   -B1 to use white background. If it is omitted, the default black background is used.
   ```
Navigating the network

4. Zoom in/out by either (i) pressing left mouse button + “z” key on the keyboard and moving the mouse, or (ii) using the mouse rotation wheel.

5. Rotate the network by pressing the left mouse button and moving the mouse. To move the whole network to your desired location in the window, press the “z” key on the keyboard and move the mouse.

6. If no problem occurs, you should get a view similar to the one in Fig. 8.27.1. Please note that all the images are captured via basic screenshot operations.

Network layout options

The current version of iCAVE includes eight different network layout algorithms (details in Liluashvili et al., 2017). Test each algorithm to see which works best in communicating your network. In Fig. 8.27.2, we illustrate different network layouts for the diseasome network using four of these algorithms.

7. Open the main menu of iCAVE by pressing the right mouse button, and scroll inside Layout Algorithms. All available layout algorithm options are listed as toggle buttons.

   i. For Fig. 8.27.2A, select the Force Directed toggle button.
   ii. For Fig. 8.27.2B, select the Coarsened Force Directed toggle button.
   iii. For Fig. 8.27.2C, select the Simulated Annealing Force Directed toggle button.
   iv. For Fig. 8.27.2D, select the Hemisphere toggle button.

8. After testing several available layouts in iCAVE (see Fig. 8.27.2A-D) for this network, we choose the weighted force directed layout as the most informative, as shown in Fig. 8.27.1. Similarly, users are encouraged to try different layouts to see which one works best for their particular network. In this protocol, we will continue performing further operations on this layout. To switch to the weighted force directed layout, go to the main menu and select Layout Algorithms → Weighted Force Directed toggle button. You should get a view similar to Fig. 8.27.1.

Adjusting size of network nodes/edges

9. To adjust node or edge size, go to main menu and select Adjust Node/Edge Size.
Figure 8.27.2 Diseasome network visualized using different layout algorithms in iCAVE. (A) Force-directed; (B) coarsened force-directed; (C) simulated annealing force-directed, (D) hemisphere layout algorithms.

Figure 8.27.3 Graphical sliders to adjust the size of nodes and edges.

10. You should get a graphical slider as shown in Fig. 8.27.3.

11. Adjust the sizes of nodes and/or edges by using the sliders.

12. Close the Size Options popup menu window by clicking on the close button on top-right.

Displaying Node Labels

13. From the main iCAVE menu, select Show Labels.

14. To interactively adjust the size of labels, go to main menu and select Adjust Label Size. Use the graphical slider to increase or decrease the size of labels.
15. Since there are a large number of nodes, you may not be able to read all network labels. For a better viewing, zoom into an area in which you are interested; now you can read all the labels. For example, you can see the region of the network comprising diseases like Dementia, Alzheimer, and Schizophrenia in the zoomed-in area of the network in Fig. 8.27.4.

16. **Optional:** Alternatively, you can modify the input file to only display labels for some nodes. In a large network such as this one, this will draw attention to certain nodes in the global network view. As an illustration, the input file (diseasome-maindiseases.txt, see Internet Resources) provides node labels only for the top 13 diseases in the network.
   i. Copy the input file diseasome-maindiseases.txt into $iCAVE_HOME/vrnet/importFiles.
   ii. Open another terminal window and change your directory to $iCAVE_HOME/vrnet.
   iii. Launch iCAVE by using the following in the command line:

```
vrnetview -Gdiseasome-maindiseases.txt -O10 -l1 -B1
```
   iv. Follow steps 13-14 to display and adjust the labels in the network.
   v. You should get a view similar to Fig. 8.27.5.

17. You can turn node labels off to view the network structure with less occlusions. For this purpose, open the main iCAVE menu, and uncheck the Show Labels toggle button. You should get a global network view similar to Fig. 8.27.1.

**Network general measures**

You can use iCAVE to calculate general network measures as summarized in Fig. 8.27.6.

18. From the main iCAVE menu, select and enable Show General Measures toggle.

19. You should get a view similar to Fig. 8.27.6. If you want to save the measures displayed in the window, click the Save Diagrams option, and several text and graph files on different network measures will be saved in your $iCAVE_HOME/vrnet/diagrams folder.
Saving Animations

Using 2D snapshots of a network is not ideal, as it projects the 3D network to a lower dimension, decreasing its communication efficiency. Therefore, iCAVE provides functionality to generate and save 3D animation videos for a rotating model of the network. Users can save and share these animations to communicate their networks with others:

20. From the main iCAVE menu, select and enable Rotate Model toggle.
21. Use the graphical sliders in the dialog box to adjust rotation speed and direction.
22. Click the Save Rotation as Movie button. Recording starts as soon as the button is clicked. Animation file (see Video 1) will be saved in your $iCAVE_HOME/vrnet folder and named animation.gif by default.
VISUALIZING MULTIPLE LAYERS OF INFORMATION IN iCAVE

Systems-level studies often pool together multiple types of information in efforts to understand their interactions within the integrated system (Butcher, Berg, & Kunkel, 2004). For example, the diseasome network discussed in Basic Protocol 1 is a good case study for exploring multiple layers of information, as it includes two different but inter-related information types: diseases and genes. In Basic Protocol 2, we guide users on how to visualize and explore the interconnections within networks that contain multiple layers of information using this diseasome network, and demonstrate how it can be easily specified in two layers in the input file. We then further introduce the edge-bundling operation, which clarifies the underlying network topology by bundling adjacent edges together. Edge bundling is advantageous when a large number of connections between network layers exist, confusing the viewer. iCAVE has several user-adjustable parameters to adjust the desired degree of granularity of the bundled edges, as summarized below.

Necessary Resources

Input File

As the input file, we will use a slightly modified version of the file used in Basic Protocol 1. This time, the disease nodes and gene nodes are specified in two separate layers (see diseasome-layers.txt, in Internet Resources). iCAVE will take the layer information into account in the visualization. Here are a few lines in the optional nodes section of the input file:

```
#nodes name label node_scale layer_code
55   Deafness  5  1
47   Leukemia  4.75828 1
...  
1329 1  2
1338 1  2
```

As you can see, we have altered two attributes in the nodes section. First, we removed the attribute color_code, as node colors in layers are automatically assigned by iCAVE, depending on the number of layers. Second, we added the attribute layer_code, where we specified all diseases in layer ‘1’, and all genes in layer ‘2’. The remaining information in the nodes section is the same as in Basic Protocol 1.

The edges section is also similar to the input file in Basic Protocol 1. The only change is on edge coloring. In this version, we will display the colors for disease-gene edges in lightskyblue, and disease-disease edges in violet. The remaining information in the edges section is the same as in Basic Protocol 1. Here are a few lines from the top and bottom of the edges section:

```
#edges node1 node2 color_code
1285  3211 lightskyblue
468   2914 lightskyblue
...  
80   366 violet
80   960 violet
```

Starting iCAVE

1. Copy the input file diseasome-layers.txt into $iCAVE_HOME/vrnet/importFiles folder.

2. Open a terminal and change your directory to $iCAVE_HOME/vrnet.
3. Launch iCAVE with the following syntax:

```
vrnetview -Gdiseasome-layers.txt -O10 -B1
```

Gdiseasome-layers.txt to open the graph specified in file
diseasome-layers.txt
-010 the file is in iCAVE format
-B1 to use white color background

4. You should get a view similar to Fig. 8.27.7. The nodes at the top layer represent
diseases, and those at the bottom layer represent genes.

**Edge bundling**

Due to large number of connections between layers, it is difficult to discern topological
properties of the network in Fig. 8.27.7. Edge bundling operation (Holten & van Wijk,
2009) in iCAVE visually bundles adjacent edges together, analogous to bundling electrical
wires or cables. Bundling is extremely useful in identifying global patterns in very large
networks. In layered networks, bundling often leads to clusters of connections between
two layers based on the connectivity patterns between the two layers. Furthermore, users
can alter the edge bundling parameter for finer or coarser clusters. Bundling the edges
of the network in Fig. 8.27.7 enables the representation of highest-density “correlation
highways” that connect diseases and genes. To perform the optional edge bundling
operation:

5. From the main menu of iCAVE, select the Network Algorithms —> Bundle Edges
button.

6. In the dialog box, choose the settings as shown in Fig. 8.27.8.

7. Click the Run button.

8. You should get a view similar to Fig. 8.27.9. Edge bundling provides a much cleaner
image and more insights regarding the global topology of the network as compared to
Fig. 8.27.7. Correlation highways between the diseases and genes can be explored fur-
ther. Also, edge bundling can be repeated with different Edge Compatibility Threshold
(ECT) values to get correlation highways at varying degrees of granularity—from
very fine to very coarse clusters.
In Basic Protocol 3, we demonstrate network topology–based clustering algorithms and cluster visualization options provided in iCave, integrated with additional edge bundling. Clustering is important in understanding the network structure, and iCave currently offers three topology-based clustering algorithms. These are edge-betweenness (Girvan & Newman, 2002), Markov (van Dongen, 2000), and modularity (Newman, 2006). In addition, iCave provides several novel algorithms for the layout of these clusters in 3D space. By default, iCave uses a force-directed algorithm for cluster layout (details in Liluashvili et al., 2017). Alternatively, users can lay out clusters using.
the 3D Circos option or display them in position, where network layout stays the same but nodes are colored based on the graph clustering results. Each algorithm has strengths and weaknesses that depend on the network under study. We recommend that users test and choose the best clustering and cluster visualization method for their network, and then add edge bundling to clarify the strength of intra-cluster and inter-cluster connectivities.

**Necessary Resources**

**Input File**

We illustrate Protocol 3 using the input file `visual-cortex.txt` (see Internet Resources), with data from Rossi & Ahmed (2015). This file specifies the network structure of the visual cortex of a mouse brain. There are 193 nodes and 214 edges. The nodes section of the input file starts with:

```plaintext
#nodes name node_color magenta3
1
2
```

The header line includes the required name attribute and the optional node_color attribute, which lets users specify a color for all nodes. The two lines below it list node names.

The optional nodes section is followed by the required edges section, which specifies all network edges. First few lines of the edges section is as follows:

```plaintext
#edges node1 node2 edge_color gray55
1 4
1 14
```

The header line includes the required attributes node1 and node2 and the optional edge_color attribute. Each line below the header specifies an individual edge.

**Starting iCAVE**

1. Copy the input file `visual-cortex.txt` into `$iCAVE_HOME/vrnet/importFiles` folder.
2. Open a terminal and change your directory to `$iCAVE_HOME/vrnet`.
3. Launch iCAVE with the following syntax:
   ```bash
   vernetview -Gvisual-cortex.txt -O10 -l1 -B1
   
   -Gvisual-cortex.txt to open the graph specified in file
   -O10 the file is in iCAVE format
   -l1 to use weighted-force directed algorithm for network layout
   -B1 to use white color background
   
   4. You should get a view similar to Fig. 8.27.10 (see Video 2 for a rotating 3D animation).

**In-position visualization**

In this approach, the current network layout (e.g., weighted force directed, simulated annealing force directed, linlog, etc.) stays the same post-clustering. As the positions of nodes are static, clustering information of nodes is communicated through node coloring. This approach is especially helpful when used in combination with layout algorithms that naturally position highly connected nodes closer (e.g., weighted force directed and simulated annealing force directed).
5. In main menu, select a clustering algorithm from the Clustering Algorithms → list options (e.g., select Clustering Algorithms —> Modularity for modularity clustering algorithm).

6. You should get a view similar to as in Fig. 8.27.11.

7. In this dialog box, choose the settings as shown in Fig. 8.27.11.

8. Click the Run button.

9. You should get a view similar to those in Fig. 8.27.12A-C. Each panel displays the same network layout with node colors corresponding to different clusters from different clustering algorithm results. These clusters visually overlap quite well with the weighted force directed algorithm used for the network layout.

**3D Circos layout visualization**

3D Circos layout is another cluster visualization algorithm. Here, nodes are positioned on the surface of a 3D hemisphere sliced into pie-like panels that correspond to each cluster.
10. After step 9 go to main menu and select Network Algorithms → Create 3D Circos button.

11. Assuming that you selected the modularity clustering algorithm in step 5, you should get a network view similar to Fig. 8.27.13A. Each cluster is in a different color on the surface of the 3D hemisphere in a pie-like panel.

edge bundling (optional)

Optionally, the network in Fig. 8.27.13A can be further refined with edge bundling to visually clarify the strength of intra-cluster and inter-cluster connectivities. For example, when there are many edges between two clusters, the bundled edges are thicker, representing stronger connectivity. The reverse is true if there are only a few edges between two clusters. The protocol steps for edge bundling are:

12. From the main menu, select the Network Algorithms → Bundle Edges button.

13. In the edge bundling dialog box, change the option for “Coloring of bundled edges” to “Use single bundling color.”

14. Click the Run button.

15. You should get a view similar to Fig. 8.27.13B, with the modularity clustering algorithm. Similarly, Fig. 8.27.14 illustrates the network view following Markov (Fig. 8.27.14A) and edge-betweenness (Fig. 8.27.14B) clustering algorithms with edge bundling. Intra-cluster and inter-cluster connectivity strengths are much more clear. Because of edge bundling, overall connectivity among clusters can be better discerned compared to individual edges drawn as in Fig. 8.27.13A. You can re-run the same operation by choosing different values for ECT and other parameters in the Edge Bundling dialog box, and compare the network layouts.

Please note that the edge bundling option is also useful in exploring the global structure of moderate-to-highly connected networks.
Figure 8.27.14 3D Circos layout of the network with edge bundling using (A) Markov clustering; (B) edge-betweenness clustering.

**Force-directed layout visualization**

Here, we illustrate when each cluster is positioned in space after the force-directed layout algorithm. Inside each cluster, nodes are organized using the hemispherical layout. As iCAVE users can select the background color in white or black, and for some network visualizations the black background provides better results, we will use a black background in this illustration.

16. Open a new terminal window and launch iCAVE with the following syntax:

```
vrnetview -Gvisual-cortex.txt -O10 -l1
```

17. From the main menu, select Clustering Algorithms —> Modularity button.

18. In this dialog box:

   i. Keep the first toggle option as Yes.
   ii. Change the second toggle option to “2-D” (this option is useful for journal figures).
   iii. Click the Run button.

   You should get a view similar to Fig. 8.27.15A. Each hemisphere is colored differently, corresponding to a different cluster. To perform edge bundling:

19. From the main menu, select Network Algorithms —> Bundle Edges button.

20. In the edge bundling dialog box, change the following settings:

   i. Edge Compatibility Threshold (ECT) value to 0.3.
   ii. Initial Step Size value to 0.4.
   iii. “Coloring of bundled edges” to “Use single bundling color.”
   iv. “Blend the coloring of bundled edges” to Yes.

21. Click the Run button.

22. You should get a view similar to Fig. 8.27.15B. Similarly, Fig. 8.27.16 illustrates the network view following Markov (Fig. 8.27.16A) and edge-betweenness (Fig. 8.27.16B) clustering algorithms (step 17) with edge bundling. Intra-cluster and inter-cluster connectivity strengths are seen in a much clearer way. You can re-run the same operation by choosing different values for ECT and other parameters in the Edge Bundling dialog box and compare the network layouts.
Figure 8.27.15  Force-directed layout of the network using modularity clustering with (A) no bundling; (B) edge bundling.

Figure 8.27.16  Force-directed layout of the network with edge bundling using (A) Markov clustering; (B) edge-betweenness clustering.
Basic Protocol 4 illustrates exploration of networks with known 3D coordinates. A wide range of biological networks represent interactions between entities that have actual 3D physical coordinates, from residue-residue interaction networks within a protein to regional connectivity networks within a brain. These can be as simple as a plain 3D model of a biological structure, or coupled with additional layers of correlation information. It can also be extremely useful to visualize multiple networks within a known 3D biological structure (Doncheva, Klein, Domingues, & Albrecht, 2011), where visual comparisons between different networks within the same structure can provide clues on the underlying data patterns. Coupled with edge bundling, these visual explorations can help provide novel insights to better identify global network structures and generate hypotheses.

**Necessary Resources**

**Input Files**

We will first visualize residue-residue binary correlations within a G-protein coupled receptor (GPCR), which is a protein that traverses through the cell membrane, transmitting signals from extracellular to intracellular space. The correlation values are calculated from molecular dynamics (MD) simulation data of Dror et al. (2013). In the input files, the edges represent pairwise correlations between the residues of a GPCR (M2 muscarinic acetylcholine receptor) for its three different states, to help better understand the effects of ligand binding to the protein.

The optional nodes section is the same for all three input files (i.e., gpcr-wildtype.txt, gpcr-orthesteric.txt, gpcr-twoligands.txt; see Internet Resources) and starts with:

```
#nodes name xyz node_color steelblue2
1 20.989 20.462 -20.321
2 23.245 23.387 -19.356
3 21.631 23.648 -15.918
```

The header line includes the required name attribute and optional xyz, and node_color attributes. The optional xyz attribute lets users specify x, y, and z coordinates of each node (as three tab-separated columns) in the input file. The optional node_color attribute lets users specify a single color, steelblue2, for all the nodes in the network.

**Protein model**

1. Copy the input files gpcr-wildtype.txt, gpcr-orthesteric.txt, and gpcr-twoligands.txt into the $iCAVE_HOME/vrnet/importFiles folder.
2. Open a terminal and change your directory to $iCAVE_HOME/vrnet.
3. Launch iCAVE with the following syntax:
   ```
   vrnetview -Ggpcr-wildtype.txt -O10 -B1
   -Ggpcr-wildtype.txt to open the graph specified in file
   gpcr-wildtype.txt
   -O10 to specify that the file is in iCAVE format
   -B1 to use white color background
   ```
4. From the main menu, select the Adjust Node/Edge Size toggle button.
5. Use the graphical sliders to adjust the size of nodes and edges.
6. Close the Size Options popup menu window by clicking on the top-right close button.
7. From the main menu of iCAVE, select Network Algorithms —> Bundle Edges button.
8. In the dialog box, use the graphical sliders and change:
   i. Edge Compatibility Threshold (ECT) value to 0.7.
   ii. Initial Step Size value to 0.1.
9. Click the Run button.
10. You should get the network view similar to Fig. 8.27.17A.
11. Open a new terminal window and change your directory to $iCAVE_HOME/vrnet.
12. Launch iCAVE with the following syntax:
    vrnetview -Ggpcr-orthosteric.txt -O10 -B1
13. Repeat steps 4 to 9.
14. You should get the network view similar to Fig. 8.27.17B.
15. Open a new terminal window and change your directory to $iCAVE_HOME/vrnet.
16. Launch iCAVE with the following syntax:
    vrnetview -Ggpcr-twoligands.txt -O10 -B1
17. Repeat steps 4 to 9.
18. You should get the network view similar to Fig. 8.27.17C.

Comparison of these side-by-side figures reveals that ligand binding to the GPCR increases correlations between residue pairs.

Next, we provide examples on visualizing a brain connectome and a neuron. Both input files (brain.txt; neuron.txt; see Internet Resources) have formats similar to the protein network input files.

**Brain model**

Brain connectome networks are widely utilized for both research and clinical purposes, and there are a large number of repositories that house such networks. For this protocol, we selected a human brain connectome generated by the PIT Bioinformatics Group (Budapest Reference Connectome Server v2.0, 2015) using MRI scans obtained from the Human Connectome Project (McNab et al., 2013).
19. Copy the input file `brain.txt` into the `$iCAVE_HOME/vrnet/importFiles` folder.

20. Open a terminal and change your directory to `$iCAVE_HOME/vrnet`.

21. Launch iCAVE with the following syntax:
   ```
   vrnetview -Gbrain.txt -O10 -B1
   -Gbrain.txt to open the graph specified in file `brain.txt`
   -O10 the file is in iCAVE format
   -B1 to use white color background
   ```

22. From the main iCAVE menu, select Adjust Node/Edge Size toggle button.

23. Use the graphical sliders to adjust the size of nodes and edges.

24. Close the Size Options popup menu window.

25. You should get the network view similar to Fig. 8.27.18.

**Neuron model**

For this model, we first selected network data of human neuron morphology (Varga, Tamas, Barzo, Olah, & Somogyi, 2015) from NeuroMorpho.Org (Ascoli, Donohue, & Halavi, 2007), which provides a large collection of publicly accessible 3D neuronal reconstructions data for the neuroscience community. We then converted the file into iCAVE input format.

26. Copy input file `neuron.txt` into `$iCAVE_HOME/vrnet/importFiles` folder.

27. Open a terminal and change your directory to `$iCAVE_HOME/vrnet`.

28. Launch iCAVE with the following syntax:
   ```
   vrnetview -Gneuron.txt -O10 -B1
   -Gneuron.txt to open the graph specified in file `neuron.txt`
   ```
Figure 8.27.19  View of the human neuron reconstruction model.

-010 the file is in iCAVE format
-B1 to use white color background

29. From the main menu, select and activate the Show Single Nodes toggle button.
30. From the main menu, select Adjust Node/Edge Size toggle button.
31. Use the graphical sliders to adjust the size of nodes and edges.
32. Close the Size Options popup menu window.
33. You should get the network view similar to Fig. 8.27.19.
INSTALLING iCAVE

iCAVE was developed using C++ and uses Vrui library based on OpenGL API platform. Before installing iCAVE, we require successful prior installation of Vrui and the libraries listed below.

Necessary Resources

Software

Operating System (OS): Linux, UNIX, MacOS
In addition, iCAVE requires that XC ode, XQuartz, libjpeg, libz, and libpng be already installed prior to iCAVE installation in MacOS platform.

1. Download the latest version of iCAVE from http://research.mssm.edu/gumuslab/software.html. Please note that as we constantly add new features, all user-ready future releases of iCAVE will be freely available for download in this Web site, and we recommend checking it periodically for newer releases.

2. Open a terminal window.

3. Change your working directory to the location where you saved iCAVE.

4. Open up the compressed package:

   ```
   tar -xvzf $iCAVE_PACKAGE_NAME
   ```

5. Change your working directory to $iCAVE_HOME.

6. Start the installation process by running:

   ```
   ./install.sh
   ```

7. If the installation process finishes successfully, proceed to step 8. If there are any errors, please check the Gümüs lab Web site (http://research.mssm.edu/gumuslab/software.html.) for possible solutions.

8. After the successful installation of iCAVE, post-installation configuration tasks need to be performed. To achieve this, run:

   ```
   source setup.bash
   ```

9. To test the iCAVE installation:
   
   i. Open your X Window terminal.
   
   ii. Make sure the necessary configurations are performed in your current terminal by running:

      ```
      source setup.bash
      ```

   iii. Change your working directory to $iCAVE_HOME/vrnet.
   
   iv. Launch iCAVE with a sample network input file with (sample file comes in the tar package):

      ```
      vrnetview -Gsample.txt -O10
      ```

10. If you get an X Window terminal with the network visualization, you can start interactive explorations of the sample network with iCAVE. If you get any error message(s), please check the Troubleshooting section for possible solutions.
NECESSARY RESOURCES FOR STEREOSCOPIC 3D OPTION AND HOW TO TURN IT ON OR OFF

iCAVE is portable between 3D, stereoscopic 3D, and immersive 3D environments. In this protocol, we describe the necessary hardware and software resources for 3D stereoscopic viewing, and the steps on how to turn on or off the stereo option.

**Necessary Resources**

**Hardware**

Users can explore stereoscopic 3D visualizations using iCAVE if they have stereo-enabled hardware and active shutter glasses that convey stereoscopic images as described below. These glasses are significantly cheaper than recent head-mounted displays (HMDs). Alternatively, users can connect their stereo-enabled computers to large screens or curved walls, which provide the additional benefit of magnification and wide field of view, or interact with their networks in immersive CAVE environments. However, we would like to note that even without any of these resources, investigators can use iCAVE to interactively explore networks in 3D using their personal computers and utilize most its functionalities.

**Stereo-enabled graphics card:** Stereoscopic 3D visualization on a 2D computer screen involves the generation of slightly different images for each eye, creating the perception of depth for the viewer. iCAVE supports professional-level quad-buffered stereo, which most modern computers come equipped with. If not, the computer can be upgraded with a compatible graphics card (e.g., Nvidia Quadro).

**Active shutter glasses:** Professional-level (active) stereo 3D viewer technology employs active shutter systems, which operate by presenting the image intended for the left eye while blocking the right eye, then presenting the right-eye image while blocking the left eye. This operation is repeated very rapidly so that the fusion of the two images is perceived as a single 3D image. Most modern active shutter 3D systems use liquid crystal (LC) shutter glasses (e.g., Nvidia 3D Vision) that are available at a sub-price range.

**120 Hz Monitor:** Due to the rapid switch between left-right eye view required by the active shutter systems, we recommend a monitor that can display at high refresh rates to prevent flickers, with a refresh rate of at least 100 Hz. Most modern LCD monitors support this refresh rate.

**Software**

**Operating System:** iCAVE runs in Mac and Linux systems for all its 3D functionalities. However, the stereo option requires a Linux-based operating system, as current Macs do not support quad-buffered stereo (yet).

Please note that iCAVE supports more basic (less restrictive; but lower quality) anaglyph stereo 3D option as well.

1. To turn on the stereo option:
   - i. Open a terminal window.
   - ii. Change your working directory to $iCAVE_HOME.
   - iii. Run the command:
     ```sh
tereo.sh
```
2. To turn off the stereo option:
   - i. Open a terminal window.
   - ii. Change your working directory to $iCAVE_HOME.
   - iii. Run the command:
     ```sh
no-stereo.sh
```
GUIDELINES FOR UNDERSTANDING RESULTS

The objective of iCAVE is to assist researchers in visualizing their networks to identify the unique characteristics of data that underlie large, dense or complex biological networks in a simple and user-intuitive fashion. Visualization of abstract networks in 3D is a nascent field, and how best to utilize features specific to 3D or to take advantage of the recent 3D technologies is currently an open research question. iCAVE can therefore also assist researchers in visualization design to perform user studies in order to better understand the relative advantages of its different 3D features. The protocols described here demonstrate different features of iCAVE for interactively exploring network visualizations in 3D, stereoscopic 3D and immersive 3D. To identify unique characteristics of the data that underlie large, dense, or complex biological networks, researchers can build on these basic protocols and utilize the full range of customization options to visually explore their networks. Overall, iCAVE is arguably the first 3D, stereoscopic 3D, and immersive 3D biomolecular network visualization tool that is open source and utilizable with commercial hardware/software. The responses and requests from its users will shape its future versions and assist in identifying its ideal usage scenarios.

COMMENTARY

Background Information

Advances in computational and experimental technologies in life sciences lead to the generation of increasingly large and complex data, making it a real challenge to identify special patterns buried within them. Visually exploring the data in the form of network representations is often more user-intuitive than exploring the raw data. However, these networks are also increasing in size and complexity, and often contain multiple subtypes. For successful explorations, effective visualization tools that can accommodate such networks are needed.

We have recently developed iCAVE, an open-source network visualization tool to help researchers interactively explore large, complex, and multi-layered biological networks in 3D. In addition, iCAVE takes advantage of recent advances in 3D visualization technologies to enable users to visualize their networks in stereoscopic 3D (using an appropriate graphics card and stereo glasses) or immersive 3D (in a CAVE environment). The 3D environment provides additional navigational capabilities that let users view a complex network from multiple perspectives, and identify structures within the data that are difficult to discern in 2D. Being able to utilize the third dimension also lets users explore multi-layered networks with greater flexibility and explore networks with known physical 3D coordinates in their original spatial structure.

The network exploration and interaction functionalities within the iCAVE 3D graphical user interface offers several alternative network layouts, clustering algorithms, and cluster visualizations to provide flexibility in addressing different user needs. Multi-layered networks or networks with physical 3D constraints can be visualized accurately. In addition, edge bundling functionality can be used during various exploration stages to identify global patterns or unique modules. iCAVE provides users with plenty of additional options to customize their networks, such as color, scaling, and directionality, as summarized in Table 8.27.1 and Table 8.27.2.

Based on the continued interest from the research community in utilizing iCAVE for various biological problems, we are continuously enhancing its capabilities. Therefore, we accompany this protocol with an updated version of iCAVE, which includes these additional functionalities:

- Addition of ‘label’ node attribute. Users can now specify different labels for all or a subset of the nodes in their network. This feature is especially helpful when highlighting a small number of important nodes within a complex network.
- Clustering dialog box, which enables users to easily choose a cluster layout option.
- Edge-bundling dialog box, for users to adjust edge-bundling parameters (e.g., granularity) and specify different options for the visualization of the bundles (e.g., coloring or color blending).

The basic protocols provided here demonstrate some of the most common usage scenarios, although many other scenarios are possible. In general, network exploration is an iterative process, and, accordingly, lessons learned at various steps can be pieced together strategically to improve the exploration experience and success.
Critical Parameters and Troubleshooting

Unable to load network

Symptom 1: When loading a network from an input file, you get an error message and iCA VE aborts the launch process as it displays an error message similar to the following:

```
libc++abi.dylib: terminating with uncaught exception of type IOError:
IO::StandardFile: Unable to open file inputFiles/fileXYZ for reading due to error 2
```

Possible causes: iCA VE cannot locate the file that was passed as an argument.

Remedies: Check the location and name of the file in your machine and make sure it matches the argument passed to iCA VE.

Symptom 2: When loading a network from an input file, you get an error message and iCA VE aborts the launch process as it displays an error message similar to the following:

```
edgecolor does not match to any known properties. Check the syntax and users manual
```

Possible causes: Input file was not generated accordingly with the iCA VE format specifications. For instance, the wrong keyword was used.

Remedies: Check the error message and make sure that your input file complies with iCA VE format specifications. For instance, the error message above complains about the usage of the `edgecolor` keyword in the input file. The problem can be resolved by changing it to the proper keyword, which is `edge_color`.

Takes long time to load network

Symptoms: When loading a network from an input file, it takes a long time to launch the iCA VE graphical interface and display the initial network view without any error messages.

Possible causes: Network consists of large number of elements (nodes, edges) and/or the choice of layout algorithm may be taking a long time. Also, it is possible that the optional annotations (e.g., labels) in the input file increase the processing time.

Remedies: Change the network layout option used during the launch of iCA VE. Some layout algorithms take a longer time than others to calculate positioning of nodes, which may delay the loading process. Try launching iCA VE with a less compute-intensive layout algorithm, such as the hemisphere layout (argument -12), or semantic levels (argument -13) algorithm.

If the `label` keyword is not used in the network specification file, iCA VE uses name of all the nodes listed in the input file to use as node labels. During launch time, iCA VE generates all the necessary labels and makes them ready to use. This process can take a while if a large number of labels need to be generated. To speed up this process (and launch time), try to limit the number of node labels specified for your network using the `label` keyword in accordance with iCA VE formatting guidelines.

Takes long time to perform an operation

Symptoms: After an operation is requested in iCA VE, the (visual) results don’t appear for a long time and the program seems frozen.

Possible causes: Requested operation requires too much computation time.

Remedies: Some functionalities provided by iCA VE require considerable amount of computation, especially for networks with large number of nodes and/or edges. In some cases, adjusting the choice of input parameters can reduce the computation time. For instance, edge bundling is a computationally intensive operation in networks with large number of edges. In the edge bundling dialog window, increasing the Edge Compatibility Threshold (ECT) parameter value and/or lowering the value of other parameters result in less computation time to complete the operation. If the results are not satisfactory, this can be repeated multiple times as necessary, and parameter values can be fine-tuned to balance the quality of results and computation time.

Videos

All video files discussed in this article can only be accessed from the online version of this article.

Video 1 diseasome.gif: Rotating 3D animation of the diseasome network.

Video 2 visual-cortex.gif: Rotating 3D animation of the mouse visual-cortex network.

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Exploring Biological Networks in 3D, Stereoscopic 3D and Immersive 3D with iCAVE

8.27.26

Supplement 61

Current Protocols in Bioinformatics

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Internet Resources

https://research.mssm.edu/gumuslab/icave/cpbi-files.

Network input files referred in this manuscript can be found here.