

Hybrid Line-Based and Region-Based Interactive Set Data Visualization

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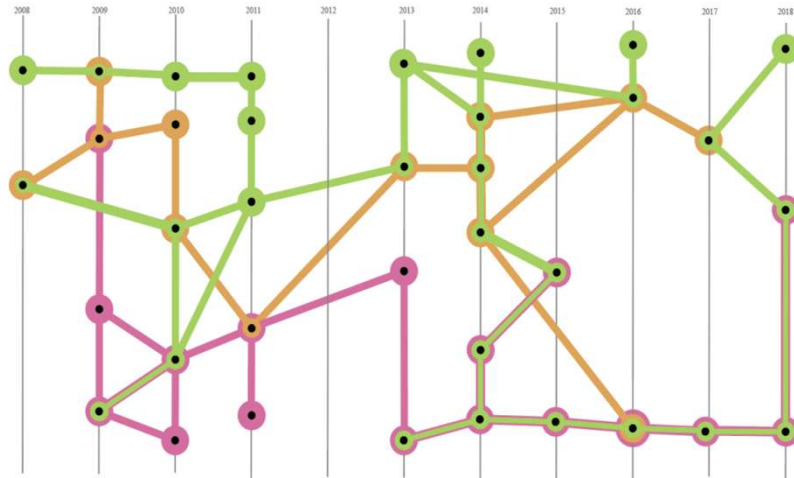
Motivation



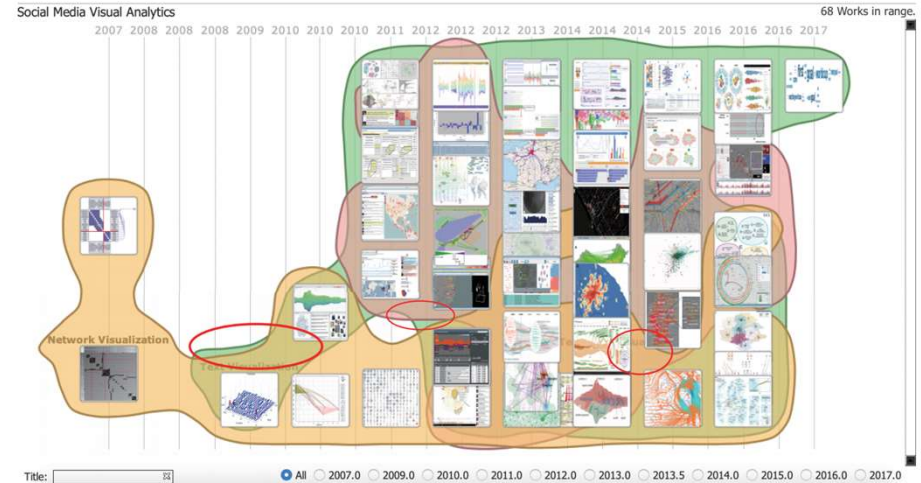
- Data exploration is popular:
 - It is challenging to visualize all set data items
- Existing methods are imperfect:
 - The **tabular** form
 - The **line-based** scheme
 - The **region-based** scheme

Reference	Year	Problem	Method	Learning	Subcategory	Steps	Visual	Code
De Bontempere et al. [22]	2015	Destination Prediction	GPs, Metadata	RL	MSP / MLP / SVM / MLP / Memory	Real	x	x
Li et al. [10]	2017	Destination Prediction	GPs, Metadata	RL	MLP	Real	x	x
Kato et al. [23]	2017	Destination Prediction	GPs	RL	RL	Real	x	x
Nayak et al. [24]	2018	Destination Prediction	Location, Metadata	RL	LSTM, SVM	Real	x	x
Ng et al. [19]	2018	Destination Prediction	Real / Real Location, Historical Demand, Metadata	RL	MLP, LSTM, LSTM	Real	x	x
Van et al. [14]	2018	Destination Prediction	Historical Demand, Graph embeddings, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [25]	2018	Destination Prediction	Historical Demand, Travel Time, Metadata	RL	CNN, LSTM	Real	x	x
Chen et al. [17]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	RL	Real	x	x
Wang et al. [26]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	MLP, CNN, LSTM	Real	x	x
Wang et al. [27]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [28]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [29]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [30]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [31]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [32]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [33]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [34]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [35]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [36]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [37]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [38]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [39]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [40]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [41]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [42]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [43]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [44]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [45]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [46]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [47]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [48]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [49]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [50]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [51]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [52]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [53]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [54]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [55]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [56]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [57]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [58]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [59]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x
Wang et al. [60]	2019	Destination Prediction	Historical Demand, Traffic, Metadata	RL	CNN, LSTM	Real	x	x

tabular



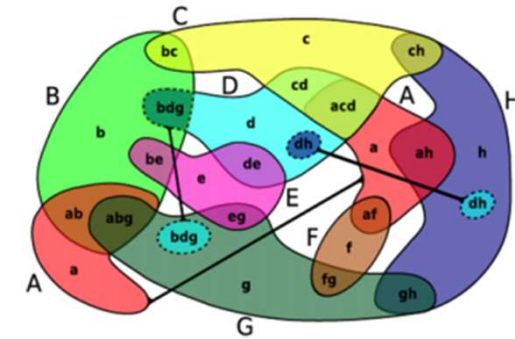
line-based



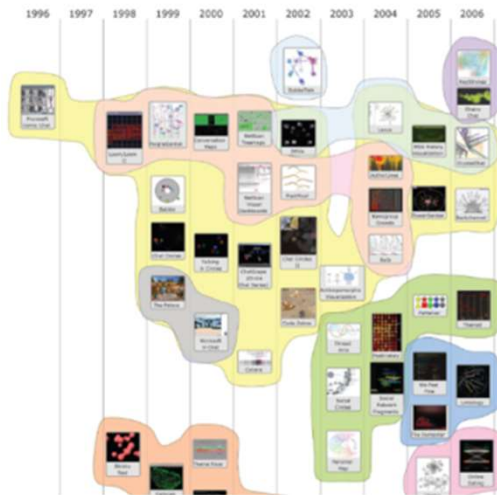
region-based

Background and Related Work

- Related Work
 - Euler and Venn diagrams
 - **Region-based** methods to visualize the set membership
 - **Line-based** methods to visualize the set membership



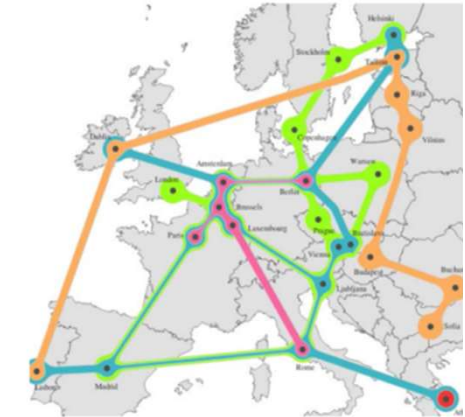
Simonetto et al



region-based overlays

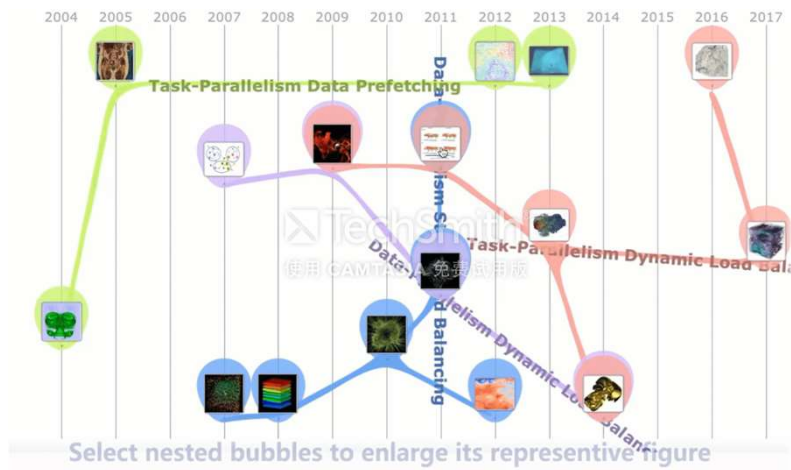


line-based overlays

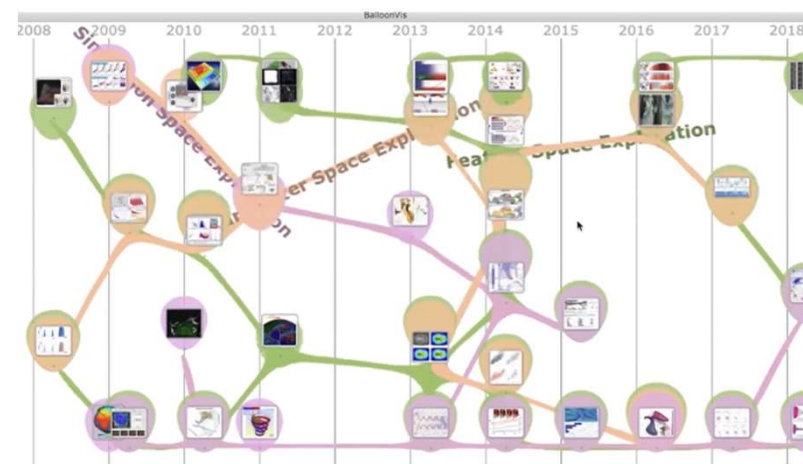


Design Rationale

- Design Goals
 - Indicate the multiple categories which the set data item belongs to clearly.
 - Allow users to directly identify set data items belonging to an identical category.
 - Alleviates artefacts caused by empty overlapping regions in region-based methods without disconnected regions.
 - Avoid too much visual clutter while preserving the original layout of the timeline.



Show representative image

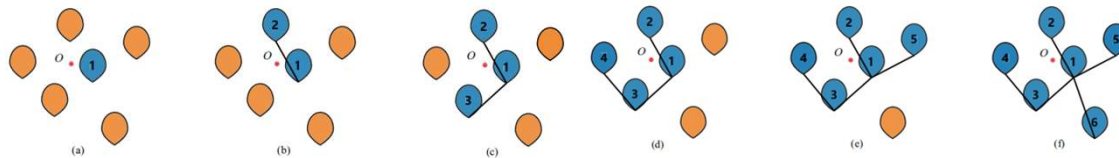


Drag items to re-layout

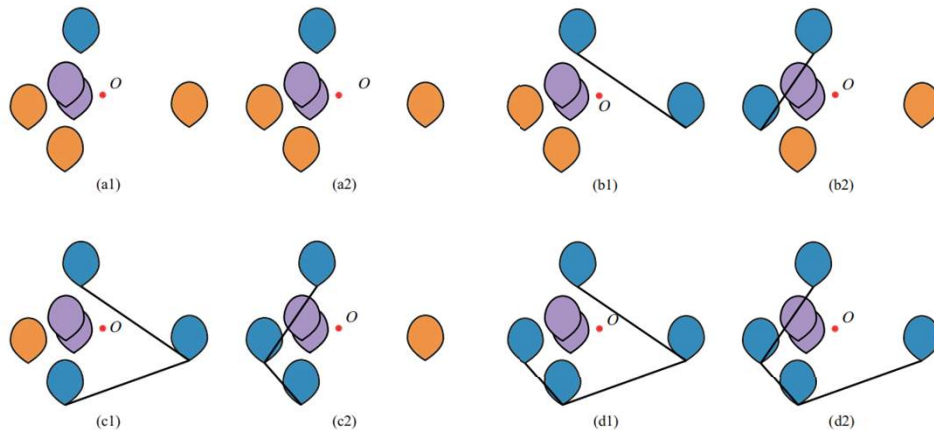
Algorithm 1: Balloon Connection



- The Connections Between Balloons



The connection process



The comparisons between the method using cost value

Algorithm 1 Balloon_Blob_Line_Connection() function.

```

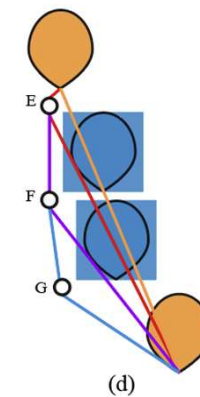
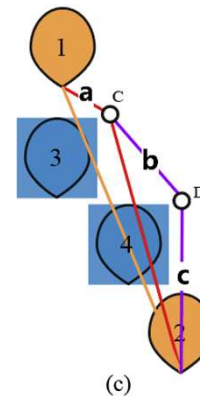
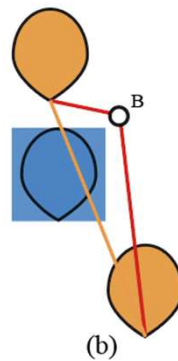
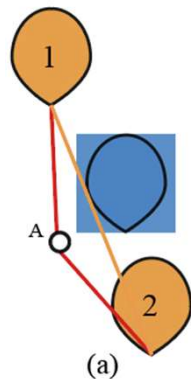
1: function BA_BLOB_LINE_CONN(start_list, end_list, sub_cato_pts[pts_size] )
2:   find central point C based on sub_cato_location[item_size]
3:   for point_iterator = 0 to pts_size do
4:     start_list.add(sub_cato_pts[point_iterator])
5:   end for
6:   for point_iterator = 0 to pts_size do
7:     cost_pts[point_iterator]=
8:        $\alpha$ *distance(start_list[point_iterator], C)
9:       +  $\beta$ *obstacles(sub_cato_pts[point_iterator], C)
10:  end for
11:  start_list.remove(item with lowest cost)
12:  end_list.add(item with lowest cost)
13:  while start_list is not empty do
14:    for start_iterator = 0 to start_list.size do
15:      for end_iterator = 0 to end_list.size do
16:        cost_pts[start_iterator][end_iterator]=
17:           $\alpha$ *distance(start_list[start_iterator],
18:            end_list[end_iterator]) +  $\beta$ *obstacles(start_list[start_iterator],
19:            end_list[end_iterator])
20:      end for
21:    end for
22:    find minimum cost_pts[M][N]
23:    Connect start_list[M] and end_list[N]
24:    start_list.remove(M)
25:    end_list.add(N)
26:  end while
27: end function
  
```

Algorithm 2: Line Connection

- Line Connection Algorithms
 - Consider each balloon as a rectangle
 - Add **control points**

Algorithm 2 Balloon_Avoidance_Line_Conn() function.

```
function BA_BLOB_LINE_CONN(start_pt, end_pt, control_pts_list)
  while The line from start_pt to end_pt intersects the rectangle do
    Find the smaller area A
    if The top left or right point is in A then
      control_pts_list.add(the top left or right point)
      start_pt=the top left or right point
    else
      control_pts_list.add(the bottom left or right point)
      start_pt=the bottom left or right point
    end if
  end while
end function
```



Algorithm 3: Hierarchical Merge



- Hierarchical Clustering and Merging
 - Different depths of the tree indicate the different degrees
 - The number of nodes is the number of balloons automatically merged into this level.

Algorithm 3 Calculate_Distance() function.

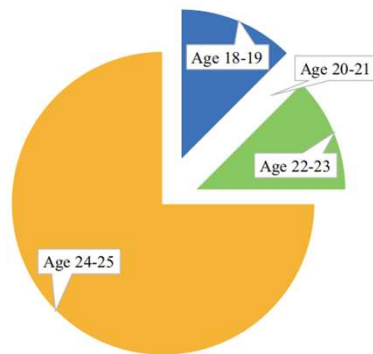
```
function CAL_DIST(possibility_cluster_a, possibility_cluster_b)
    dist ← 0
    for each grid point do
        possibility_a = possibility_cluster_a[grid point]
        possibility_b = possibility_cluster_b[grid point]
        dist ← dist + |possibility_a - possibility_b|
    end for
    return dist
end function
```

Algorithm 4 Balloon_Merge() function.

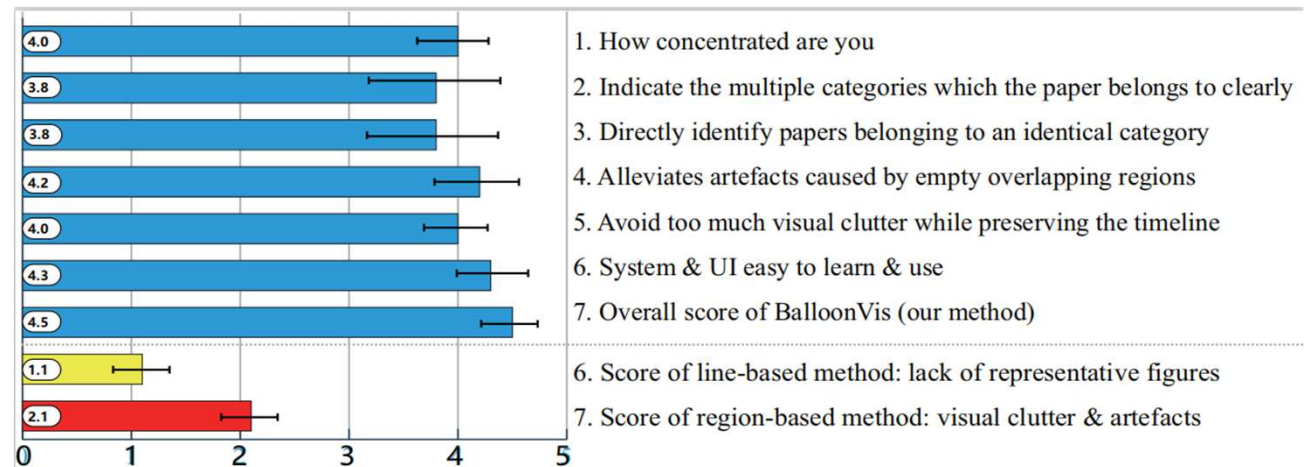
```
function BA_MERGE(ba_list, new_run_list, glyph_list)
    for ba_iterator = 0 to ba_list.size do
        cur_run_list = ba_list[ba_iterator].getRunList()
        for run_iterator = 0 to cur_run_list.size do
            if cur_is_not_new(cur_run_list[run_iterator], new_run_list) then
                new_run_list.add(cur_run_list[run_iterator])
            end if
        end for
    end for
    for ba_iterator = 0 to ba_list.size do
        glyph_list.add(ba_list[ba_iterator].getGlyph())
    end for
    new_glyph = merge_Glyph(glyph_list)
    new_balloon = construct_balloon(new_run_list, new_glyph)
    for ba_iterator = 0 to ba_list.size do
        BalloonVis.add(new_balloon)
    end for
end function
```

User Study

- We recruited **12** participants
- The participants identify literatures cited in a survey paper by:
 - Line-based method
 - Region-based method
 - Our method



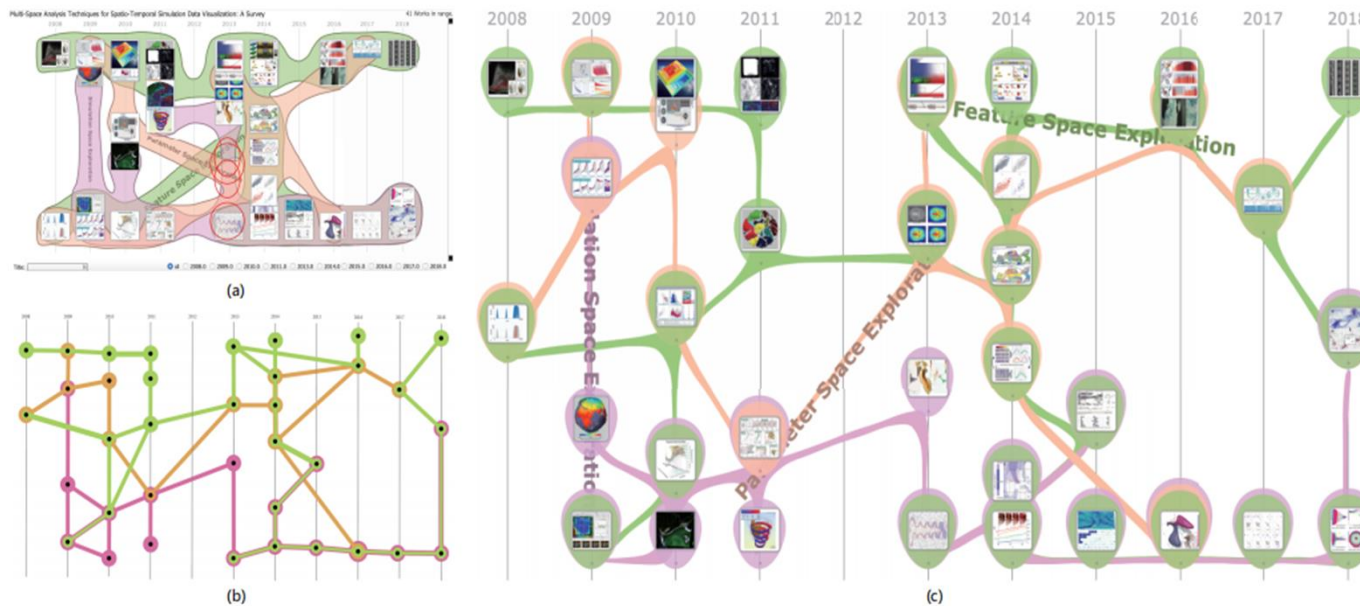
Age distribution



Post-study results

Case Study

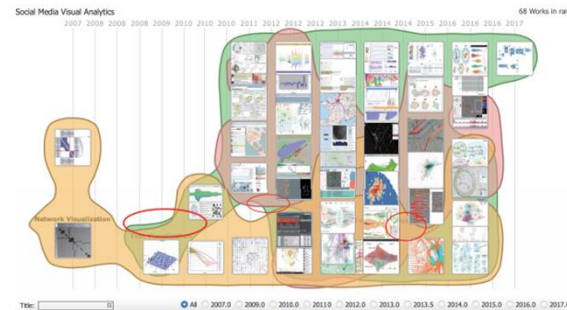
- Conduct three evaluation tests on:
 - The survey of emerging trends of deep learning methods
 - The survey on simulation data visualization
 - The survey on social media data visualization



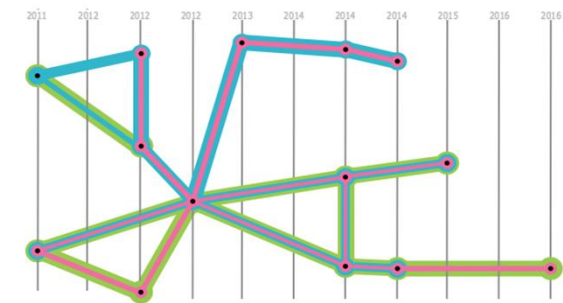
A survey paper on “Simulation data visualization”

Case Study

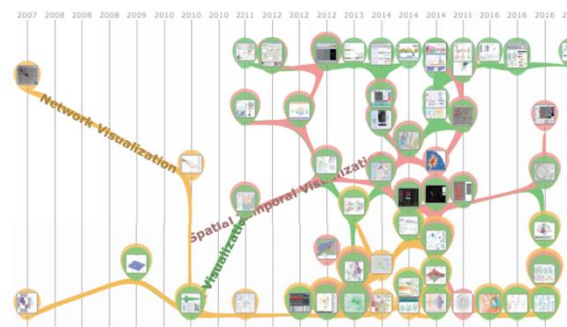
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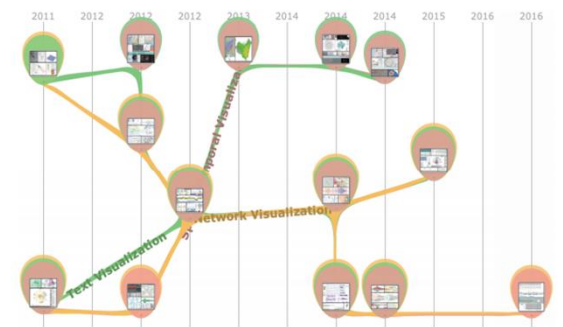
(a)



(b)



(c)

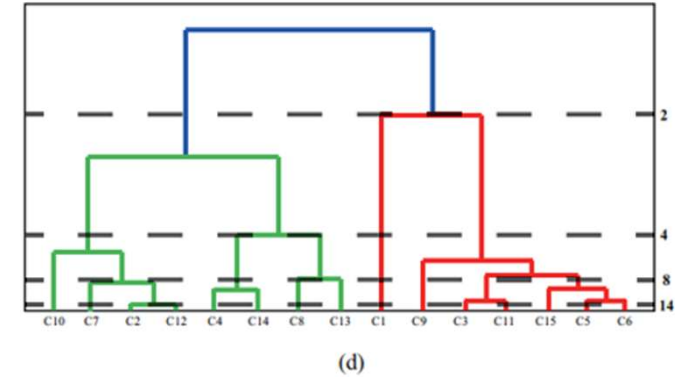
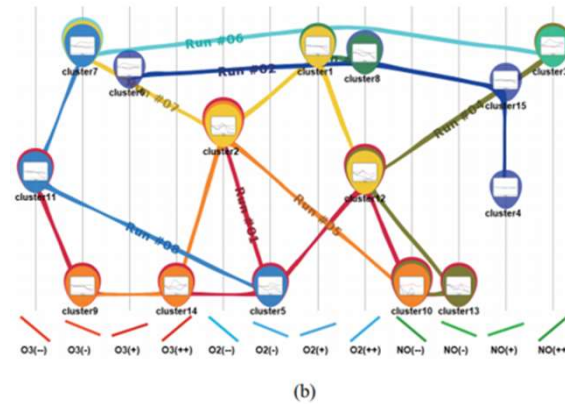
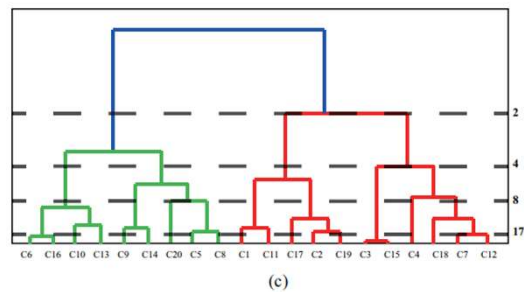
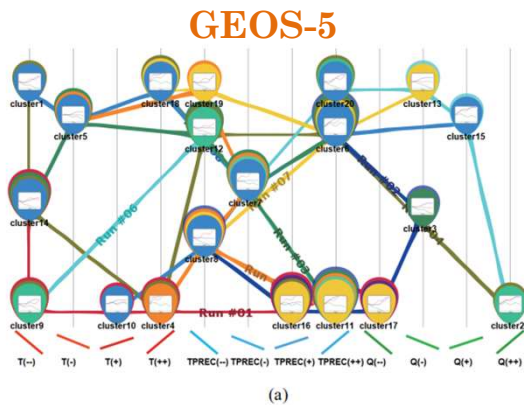


(d)

A survey on “Social media data visualization”

Case Study

- Case study on two cases:
 - Goddard Earth Observing System Model, Version 5 (GEOS-5)
 - Model of Ozone and Related Tracers, Version 4 (MOZART-4)
- Visualize **set information** in the overlapping clusters and overlapping simulation runs



MOZART-4

Discussion and Future Work

- **Limitations**
 - Introduce visual clutter when the number of data items is too large
 - It is hard to perceive the colors when there are too many set categories
- **Future Work**
 - Enable users better control the merging processes interactively when hierarchical merging

Conclusions

- A novel method to explore set relations interactively in set data based on hybrid strategy: line-based and region-based
- Support placing the representative illustration for each set data item
- Reduce visual clutters
- The scalability is guaranteed



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Thank You!