

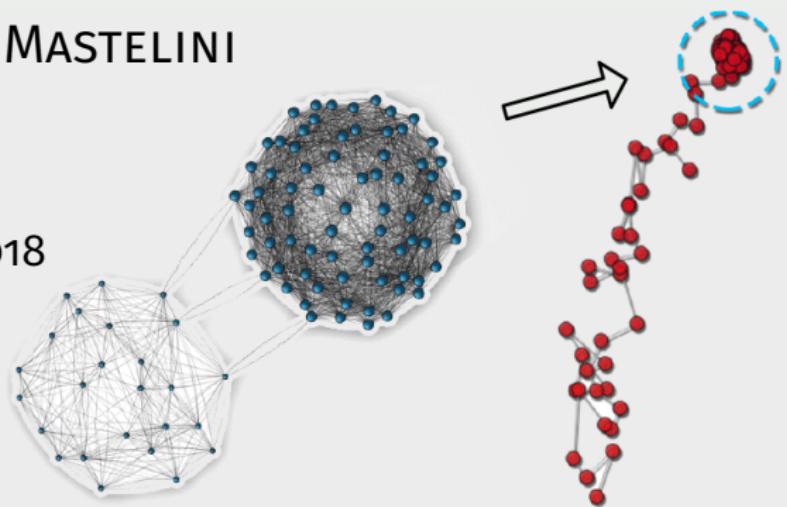
# **REDUCING SNAPSHOTS TO POINTS:**

A VISUAL ANALYTICS APPROACH TO DYNAMIC NETWORK EXPLORATION

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ICMC – USP

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# OUTLINE

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1 Introduction

2 Related Work

3 Reducing Snapshots to Points

- Discretization
- Vectorization and Normalization
- Dimensionality Reduction
- Visualization Interaction

4 Use cases

5 Discussion

6 Conclusion

# INTRODUCTION

# ARTICLE DETAILS

## Title:

- *Reducing Snapshots to Points: A Visual Analytics Approach to Dynamic Network Exploration*

## Authors:

- Stef van den Elzen
- Danny Holten
- Jorik Blaas
- Jarke J. van Wijk

## Year:

- 2016

## Journal:

- IEEE Transactions on Visualization and Computer Graphics

## Citation count:

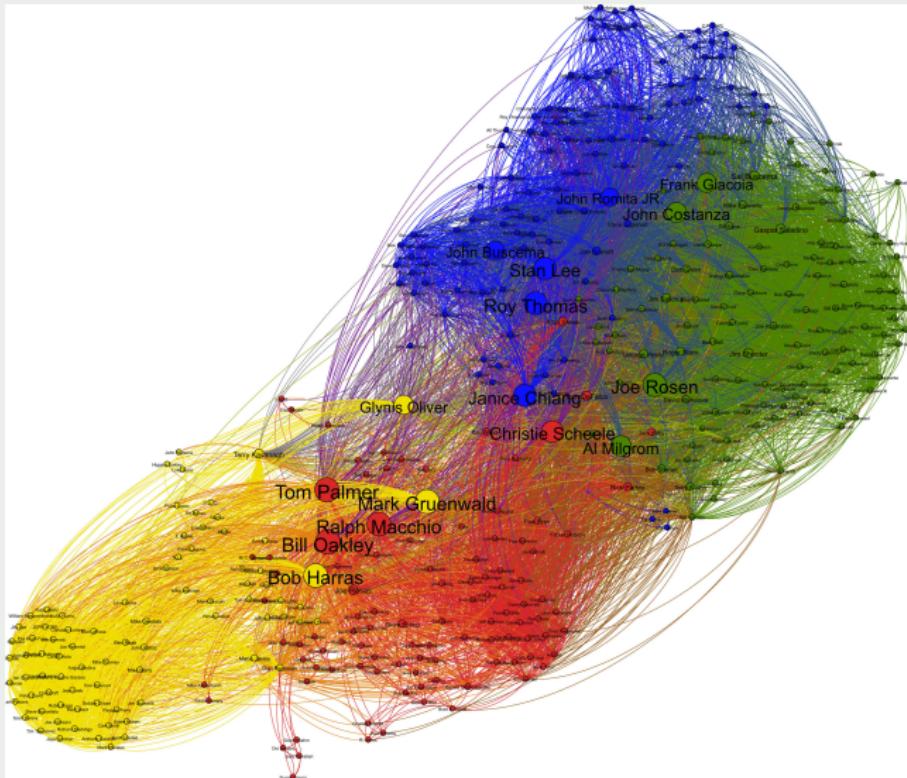
- 56
- Graph and dynamic network visualization
- Time varying visualization

# MOTIVATION

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- *Networks* are ubiquitous
  - ▶ Represent relations between objects
- Often networks are large and dynamic
  - ▶ Change over time
  - ▶ Evolve!
- Examples:
  - ▶ (Tele-) communication networks
  - ▶ Social networks
  - ▶ Financial Networks
  - ▶ Transport network

# MARVEL COMICS' AVENGERS ARTISTS COLLABORATION (1963-1996)



Available in: <http://allthingsgraphed.com/public/images/marvel/avengers.svg>

# CHALLENGES

- Understanding the evolution of dynamic networks is challenging
- Discovery of states that characterize them over time:

## *States:*

- ▶ Stable
- ▶ Recurring
- ▶ Outliers
- ▶ Transitions

## *Changes:*

- ▶ Gradual, from a state to another
- ▶ Alternated between multiple states
- ▶ Not stable at all

## TWOFOLD CHALLENGE:

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1. How to visualize, interact with, and analyze a large static network for one point in time (*snapshot*)?
2. How to visualize, interact with, and analyze many of these snapshots?

# CONTRIBUTIONS

- Proposal:** reducing network snapshots to points in 2D spaces
- Visual identification of network states
  - Identification of transitions between states
  - Analysis of network evolution in general
  
  - Visualization tool prototype as concept proof

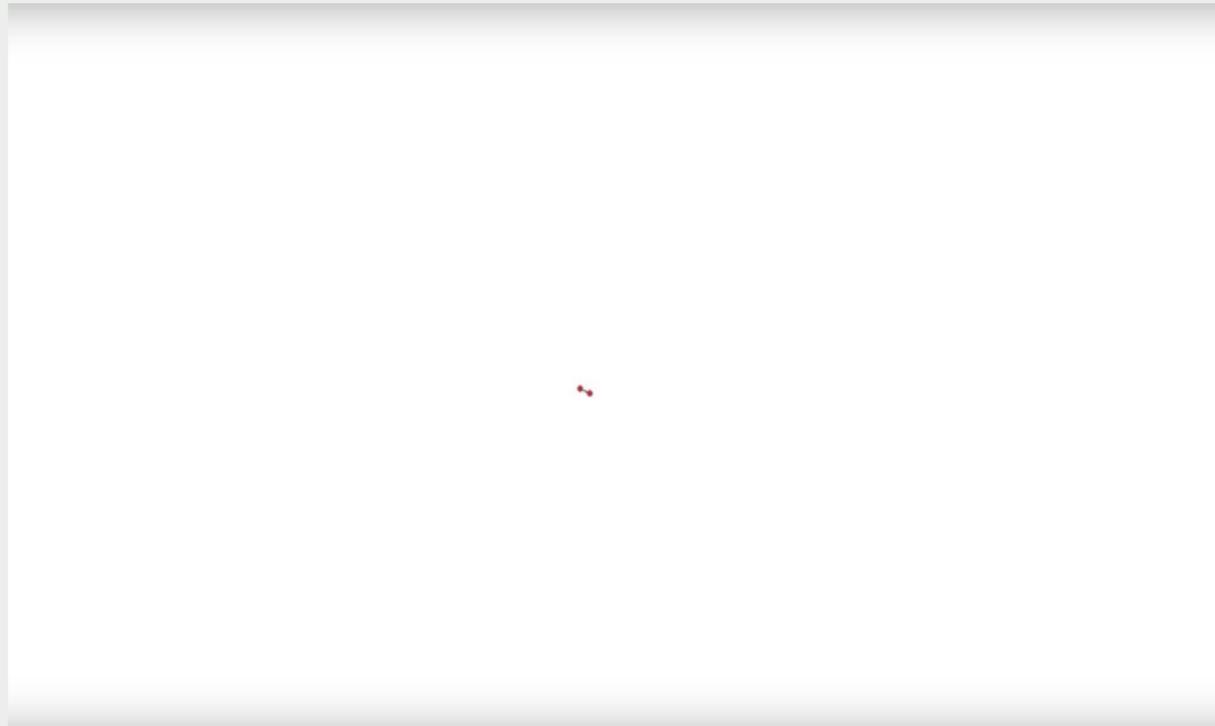
# **RELATED WORK**

# TRADITIONAL VISUALIZATION APPROACHES

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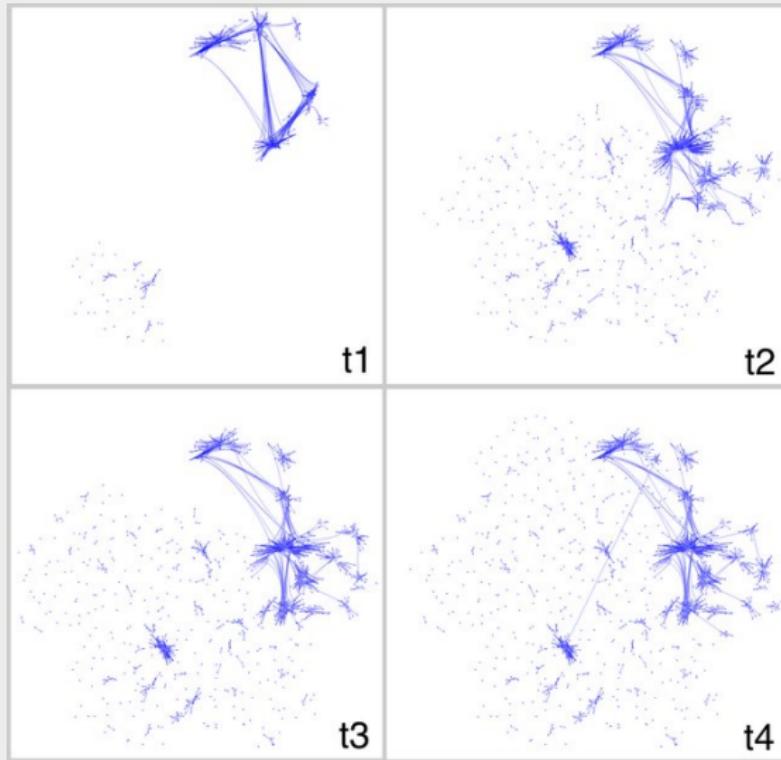
1. Mapping *time-to-time*
  - ▶ Animation
2. Mapping *time-to-space*
  - ▶ Small-multiples

# ANIMATION



Available in: <https://www.youtube.com/watch?v=HP63A09B1YM>

# SMALL-MULTIPLES



Available in: <https://arxiv.org/pdf/1409.5034.pdf>

# CURRENT APPROACHES

## ■ Problems with these approaches:

### Animations

High cognitive load:

- ▶ Difficulty to track changes over time
- ▶ How to visually encode these changes

### Small-multiples

To find the optimal balance between using:

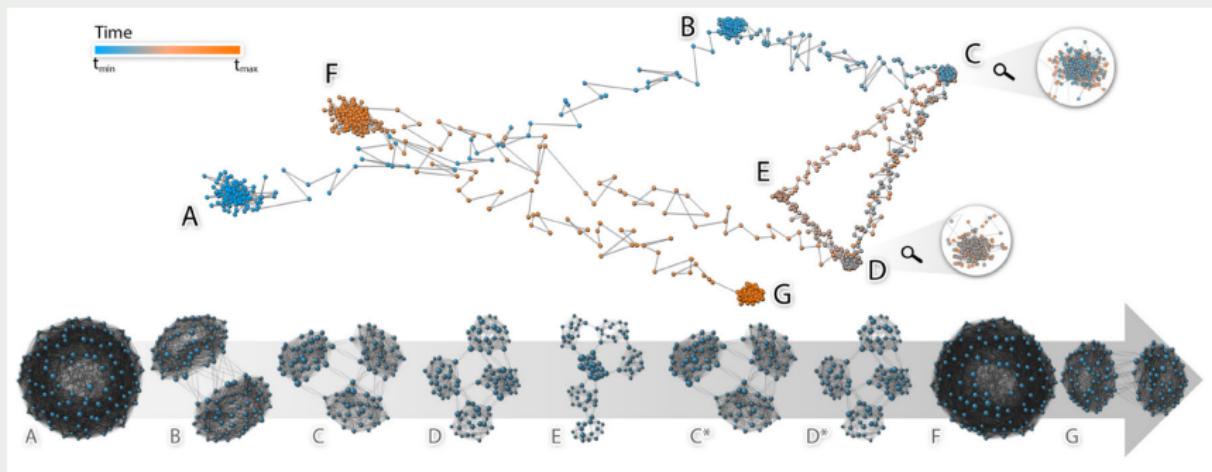
**Few images** Lacking of temporal detail

**Too many images** Difficult to interpret

# **REDUCING SNAPSHOTS TO POINTS**

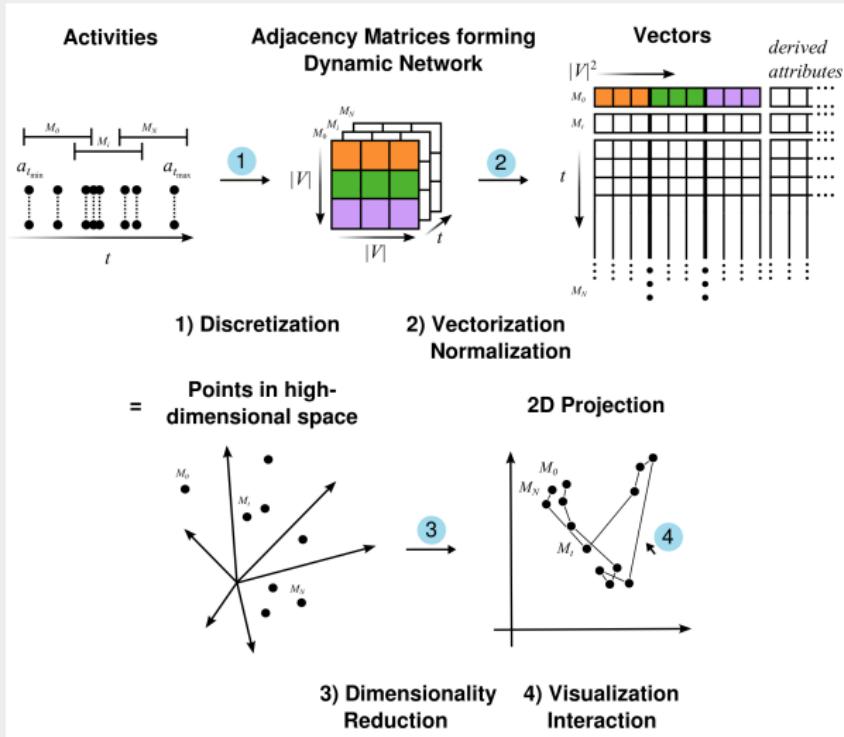
# GENERAL IDEA

Observe the disposition of projected network's snapshots,  
seeking for visible patterns



Retrieved from van den Elzen et al. [1]

## APPROACH OVERVIEW



Adapted from van den Elzen et al. [1]

# DYNAMIC NETWORK MODEL

- Dynamic Network  $\Gamma$  as a sequence of  $N$  snapshots:

$$\Gamma = (G_1, G_2, \dots, G_N)$$

- Snapshot as a directed graph:

$$G_i = (V, E_i, t_i)$$

$V$  : Node (vertex) set

$E_i \subseteq V \times V$  : Edge (link, event) set (vertex tuples  $(v_m, v_n)$ )

$t_i$  :  $i$ -th timestep

- Complete edge set and weight function  $w$

$$E = \bigcup_{i=1}^N E_i \qquad w : i \times E \rightarrow \mathbb{R}$$

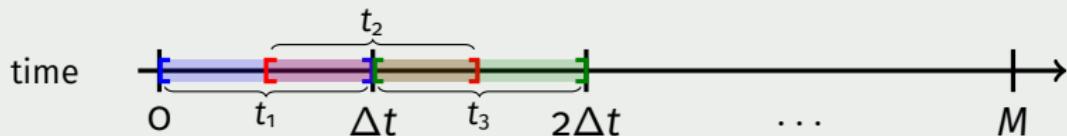
$w(i, e)$ , for edge  $e \in E_i$

# (1) DISCRETIZATION

- Discrete activities: activity log

$$A = (A_1, A_2, \dots, A_j, \dots, A_M) \quad A_j \in E \times \mathbb{R}$$

- $A_j = (a_j, s_j)$  is an activity with edge  $(v_m, v_n)$  and timestamp  $s_j$
- Time window to select snapshots with constant  $\Delta t$ 
  - ▶ Overlapping is allowed



- Each snapshot  $G_i$  is a *adjacency matrix*  $M$  of size  $|V| \times |V|$ :

$w_i$	a	b	c	...
a	$w_{a,a}$	$w_{a,b}$	$w_{a,c}$	...
b	$w_{b,a}$	$w_{b,b}$	$w_{b,c}$	...
c	$w_{c,a}$	$w_{c,b}$	$w_{c,c}$	...
...	...	...	...	...

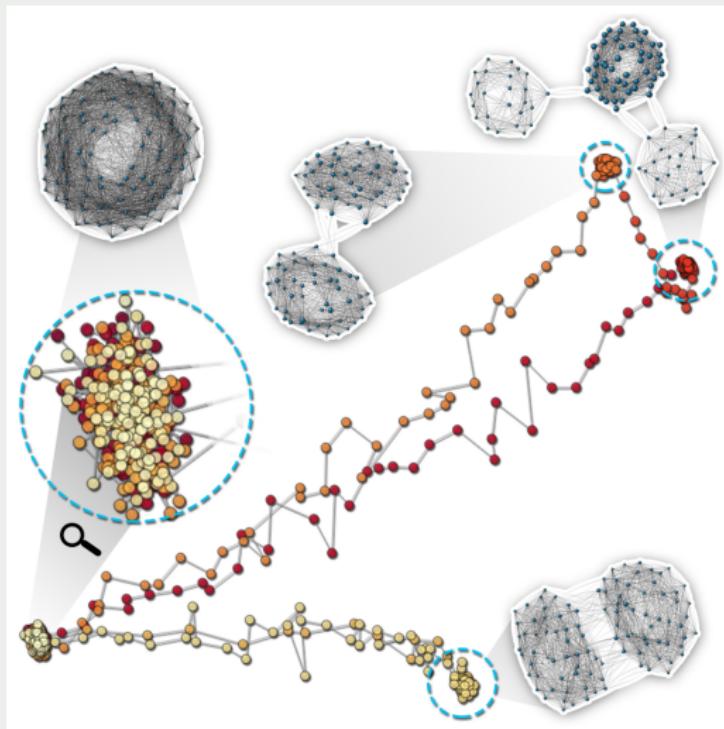
## (2) VECTORIZATION AND NORMALIZATION

- Each adjacency matrix rearranged as  $1 \times |V|^2$  array
- Snapshots stacked to form a  $N \times |V|^2$  matrix
  - ▶ Columns represent edges, rows represent snapshots
- Different normalization strategies:
  - ▶ None
  - ▶ Binarization
  - ▶ Min-max
  - ▶ Z-score
- The choice depends on the dealt problem

### (3) DIMENSIONALITY REDUCTION

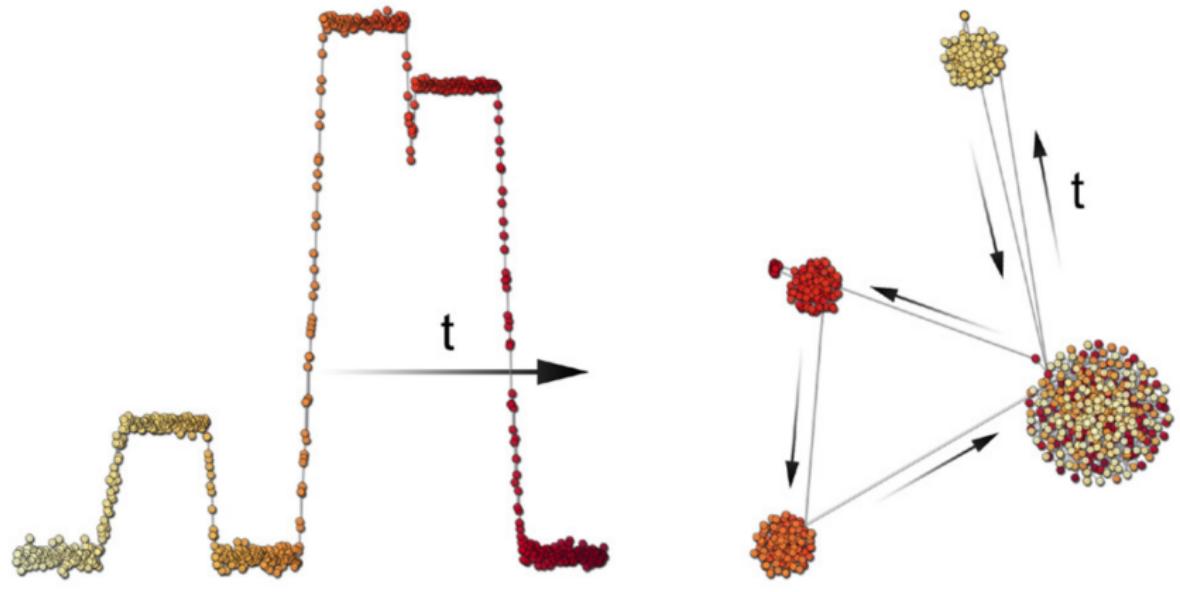
- Linear and non-linear projection techniques evaluated:
  - ▶ Principal Component Analysis (PCA)
  - ▶ t-Distributed Stochastic Neighbour Embedding (t-SNE)
- Improved versions to reduce computations:
  - ▶ *Randomized PCA*
  - ▶ *Barnes-Hut-SNE*

# EXAMPLE 1: SYNTHETIC DYNAMIC NETWORK AND PCA



Retrieved from van den Elzen et al. [1]

## EXAMPLE 2: TIME AND PC1, NON-LINEAR PROJECTION



Retrieved from van den Elzen et al. [1]

## (4) VISUALIZATION INTERACTION

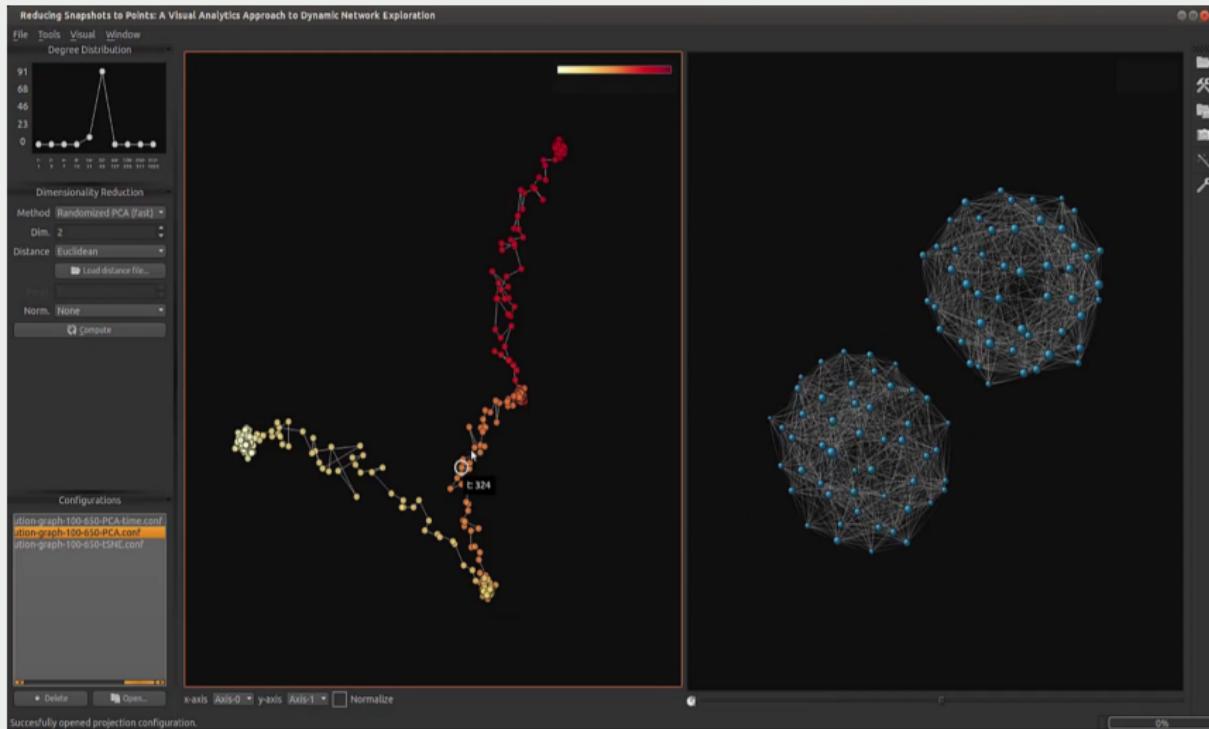
- Prototype developed as concept proof
- Two linked and juxtaposed views:

**Projection View:** Each snapshot as a dot

- ▶ Close dots represent similar network states
- ▶ Sequential dots connected with lines
- ▶ User defined colormap (e.g., global time, hour of the day)

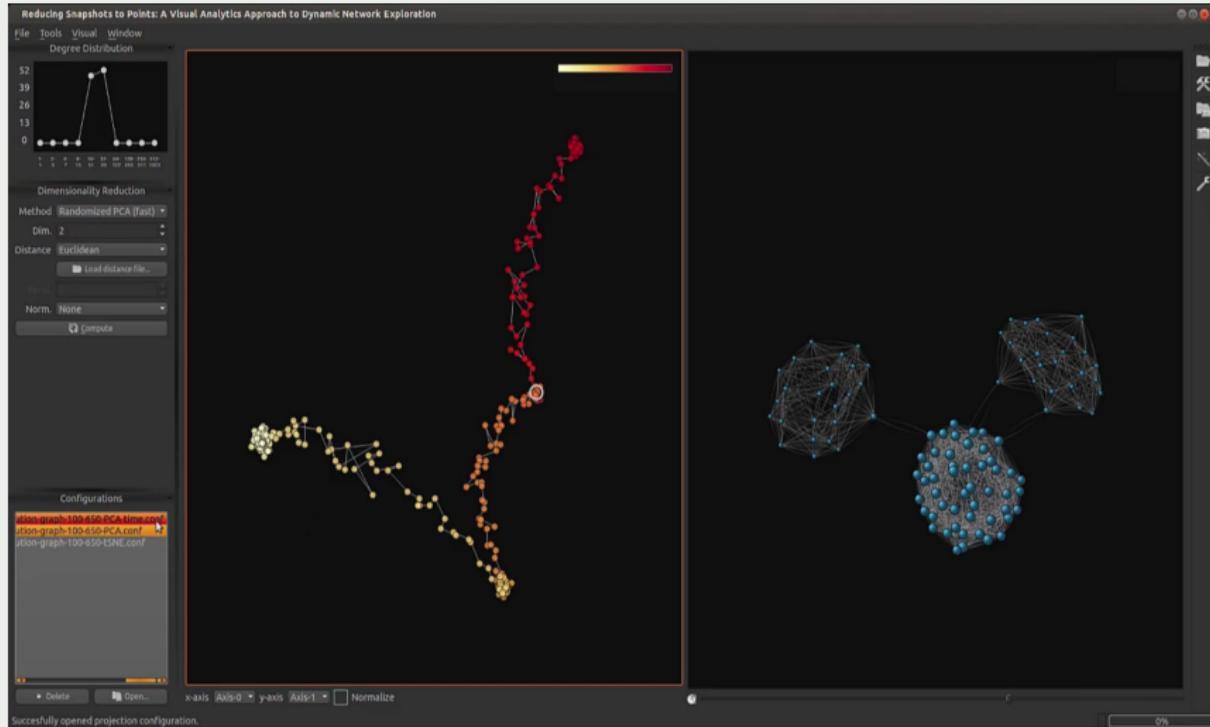
**Network View:** visualization of a selected snapshot using a node-like diagram

# EXAMPLE 3: PROJECTION VIEW AND NETWORK VIEW



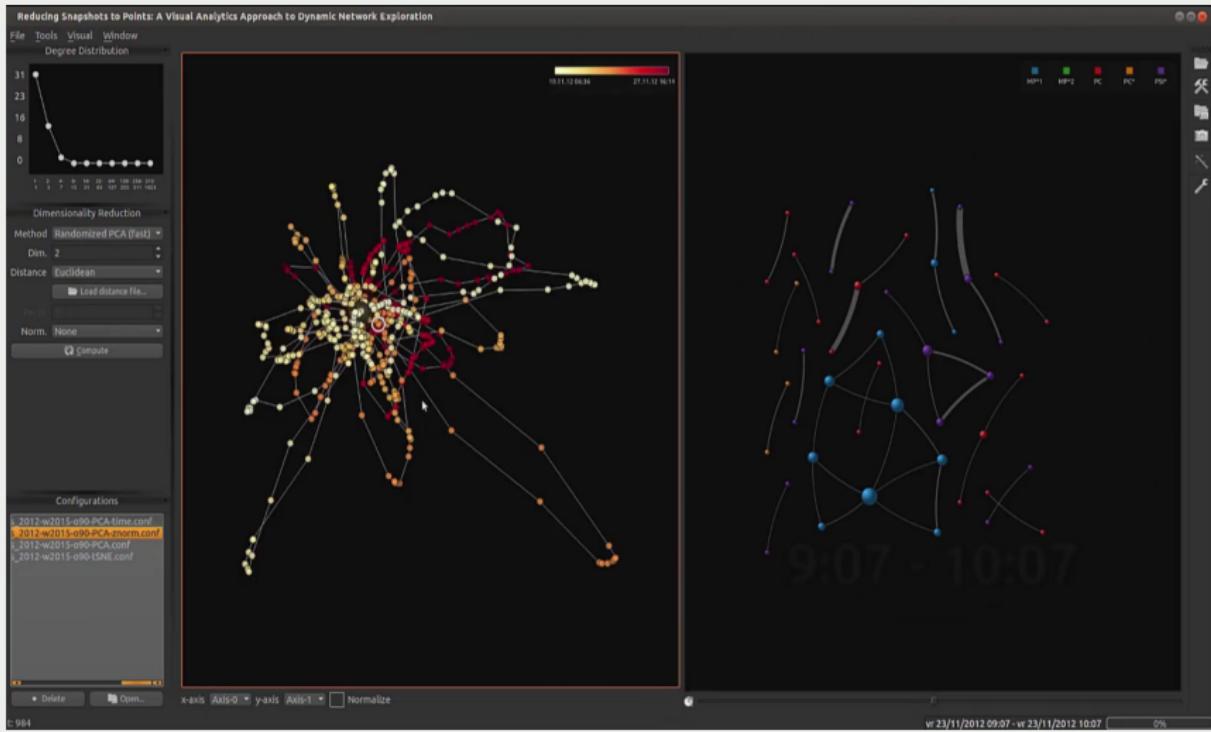
Supplementary material of van den Elzen et al. [1]

# EXAMPLE 4: DIFFERENT PROJECTIONS



Supplementary material of van den Elzen et al. [1]

# EXAMPLE 5: DEALING WITH DATA CLUTTERING AND ANIMATIONS IN THE NETWORK VIEW



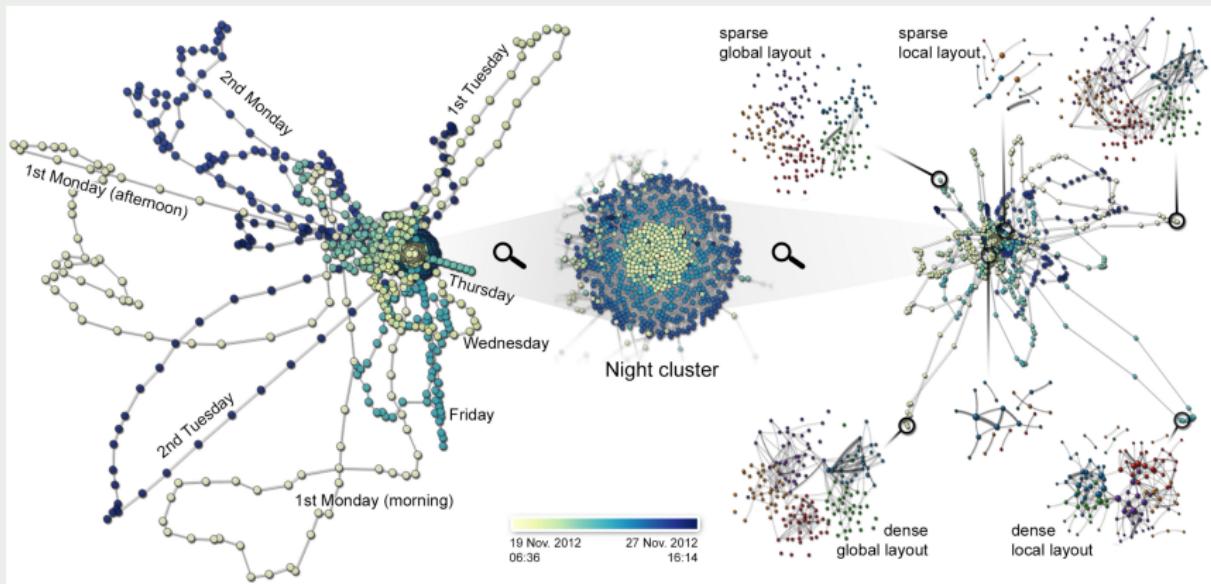
# USE CASES

# HIGH-SCHOOL CONTACT PATTERNS

- *SocioPatterns* initiative
- Face-to-face contact between persons (wearable sensors)
  - ▶ How infectious diseases spread within a population
  - ▶ High-school students
  - ▶ 7 school days (Monday to Tuesday, in a week of 2012)
  - ▶ 180 nodes (students from 5 classes)
  - ▶ 45,047 contacts (edges)
  - ▶ 10,104 unique edges
  - ▶ Time window of 60 minutes

# HIGH-SCHOOL CONTACT PATTERNS: LINEAR PROJECTION

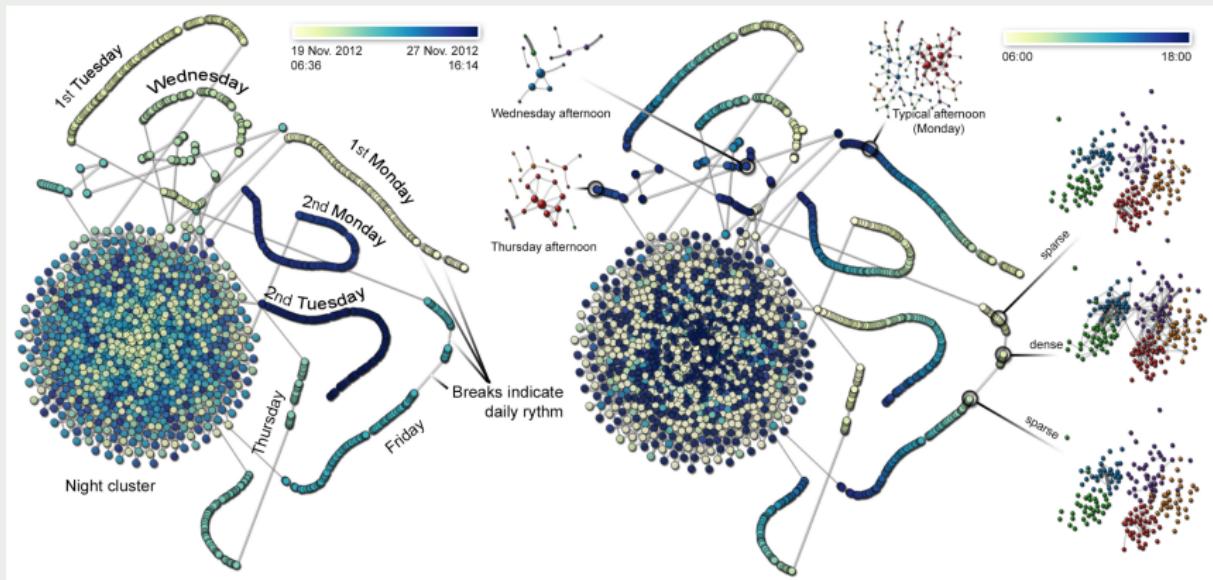
No normalization (left) and z-score (right)



Retrieved from van den Elzen et al. [1]

# HIGH-SCHOOL CONTACT PATTERNS: NON-LINEAR PROJECTION

Global time variation (left) and Hour of the day (right)



Retrieved from van den Elzen et al. [1]

# **DISCUSSION**

# DISCUSSION

- Different tools and parameters for different problems
- Two dimensional projections:
  - ▶ Linear projections show transitions
  - ▶ Non-linear projections omit transitions but are better in separating the snapshots
- Sparse vs. Dense networks
  - ▶ Distance from the center

# **CONCLUSION**

# CONCLUSION

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- Simple but yet powerful approach
- Flexibility is an advantage
- Known tools assembled in a novel approach
- Useful insights about dynamic network evolution

**Text structure and quality:** well written but sometimes repetitive/redundant

**Software availability:** prototype tool was not made available by the authors

THANKS FOR YOUR TIME!

ANY QUESTIONS?

# REFERENCES

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-  STEF VAN DEN ELZEN, DANNY HOLLEN, JORIK BLAAS, AND JARKE J VAN WIJK.  
**REDUCING SNAPSHOTS TO POINTS: A VISUAL ANALYTICS APPROACH TO DYNAMIC NETWORK EXPLORATION.**  
*IEEE transactions on visualization and computer graphics*,  
22(1):1–10, 2016.

## QUESTIONS ABOUT THE PAPER

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1. What does the agglomeration of points with mixed colors mean?
2. What type of projection achieves the best data separation? Linear or non-linear?
3. Why is z-score the indicated normalization method to be used with PCA?