



BAYESIALAB

VR

Probabilistic Latent Factor Induction With Bayesian Networks & BayesiaLab

Today's Program

Introduction

- Our Company and Technology

Motivation & Objectives

- Personality Models
- Factor Discovery

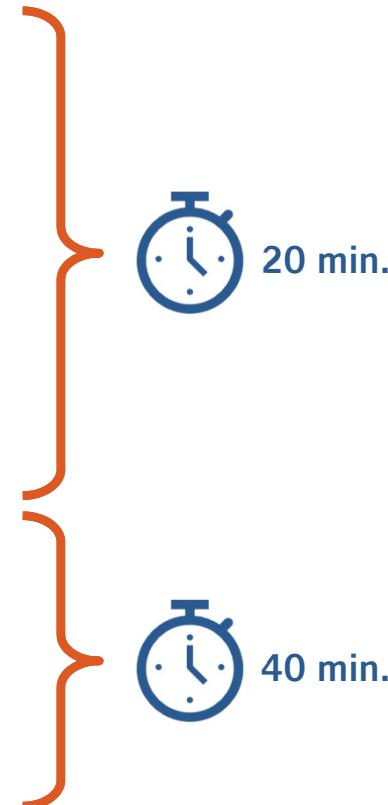
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Why Bayesian Networks?

- Nomenclature
- Arc Force

BayesiaLab Workflow

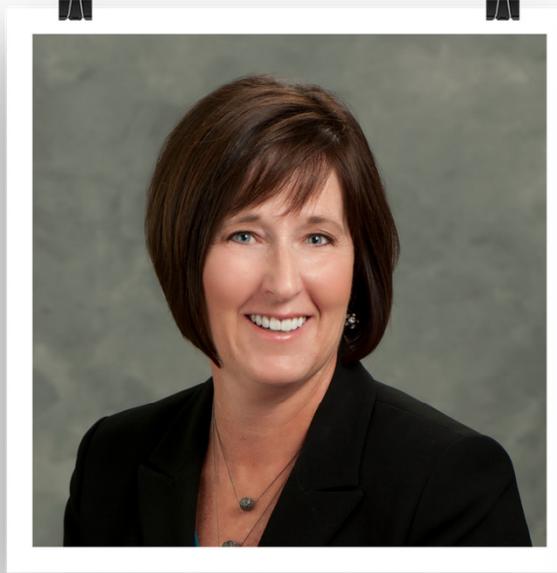
- Unsupervised Learning
- Variable Clustering
- Validation
- Multiple Clustering



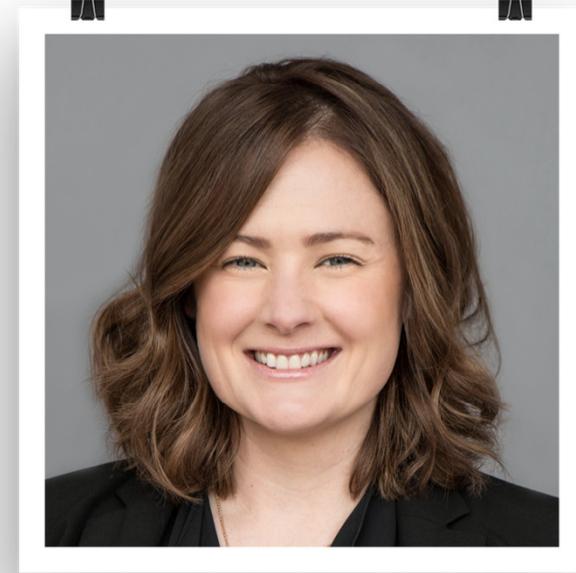
Your BayesiaLab Team Today



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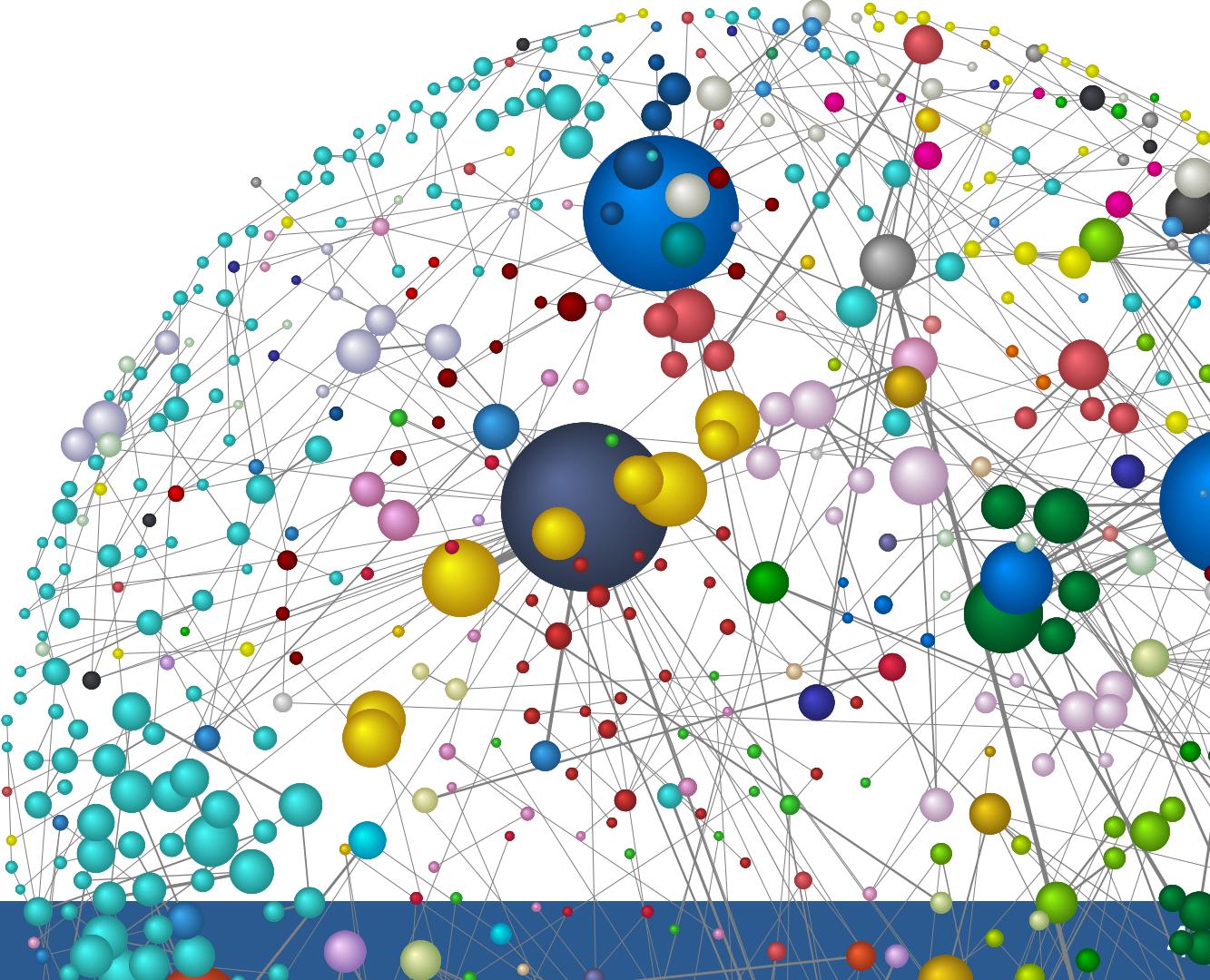
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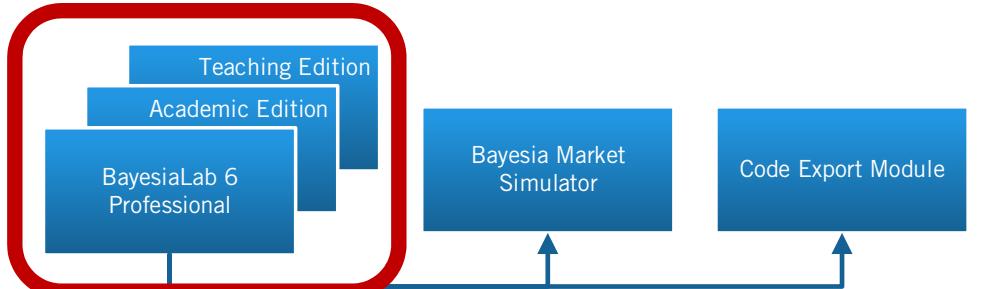
Co-founded in 2001
by Dr. Lionel Jouffe &
Dr. Paul Munteanu



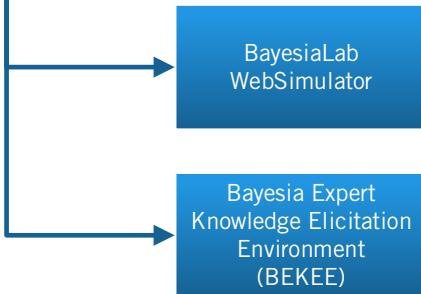
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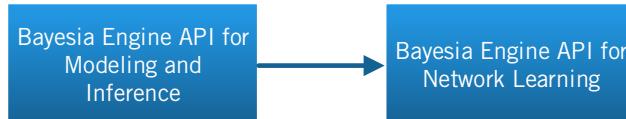
Desktop Software



Web Application



API





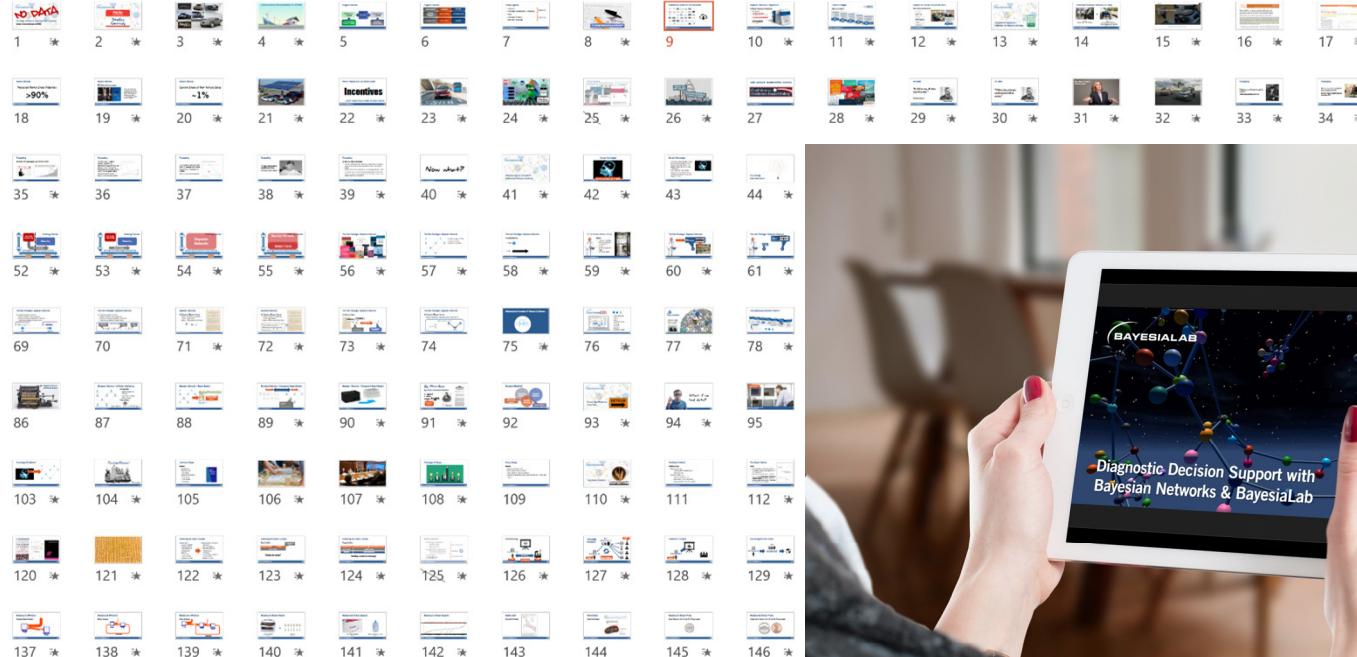
Bayesian Networks & BayesiaLab

A Practical Introduction for Researchers

- Free download:
www.bayesia.com/book
- Hardcopy available on Amazon:
<http://amzn.com/0996533303>
- See Chapter 8



Webinar Slides, Data, and Recording Available





Background & Motivation

The Dimensions of Human Personality

Accessible Active Adaptable Admirable Adventurous Agreeable Alert Allocentric Amiable Anticipative Appreciative Articulate Aspiring Athletic Attractive Balanced Benevolent Brilliant Calm Capable Captivating Caring Challenging Charismatic Charming Cheerful Clean Clear-headed Clever Colorful Companionly Compassionate Conciliatory Confident Conscientious Considerate Constant Contemplative Cooperative Courageous Courteous Creative Cultured Curious Daring Debonair Decent Decisive Dedicated Deep Dignified Directed Disciplined Discreet Dramatic Dutiful Dynamic Earnest Ebullient Educated Efficient Elegant Eloquent Empathetic Energetic Enthusiastic Esthetic Exciting Extraordinary Fair Faithful Farsighted Felicific Firm Flexible Focused Forceful Forgiving Forthright Freethinking Friendly Fun-loving Gallant Generous Gentle Genuine Good-natured Gracious Hardworking Healthy Hearty Helpful Heroic High-minded Honest Honorable Humble Humorous Idealistic Imaginative Impressive Incisive Incorrputible Independent Individualistic Innovative Inoffensive Insightful Insouciant Intelligent Intuitive Invulnerable Kind Knowledge Leaderlike Leisurely Liberal Logical Lovable Loyal Lyrical Magnanimous Manly Many-sided Masculine Mature Methodical Meticulous Moderate Modest Multi-leveled Neat Nonauthoritarian Objective Observant Open Optimistic Orderly Organized Original Painstaking Passionate Patient Patriotic Peaceful Perceptive Perfectionist Personable Persuasive Planful Playful Polished Popular Practical Precise Principled Profound Protean Protective Providential Prudent Punctual Purposeful Rational Realistic Reflective Relaxed Reliable Resourceful Respectful Responsible Responsive Reverential Romantic Rustic Sage Sane Scholarly Scrupulous Secure Selfless Self-critical Self-defacing Self-denying Self-reliant Self-sufficient Sensitive Sentimental Seraphic Serious Sexy Sharing Shrewd Simple Skillful Sober Sociable Solid Sophisticated Spontaneous Sporting Stable Steadfast Steady Stoic Strong Studious Suave Subtle Sweet Sympathetic Systematic Tasteful Teacherly Thorough Tidy Tolerant Tractable Trusting Uncomplaining Understanding Undogmatic Unfoolable Upright Urbane Venturesome Vivacious Warm Well-bred Well-read Well-rounded Winning Wise Witty Youthful Absentminded Aggressive Ambitious Amusing Artful Ascetic Authoritarian Big-thinking Boyish Breezy Businesslike Busy Casual Cerebral Chummy Circumspect Competitive Complex Confidential Conservative Contradictory Crisp Cute Deceptive Determined Dominating Dreamy Driving Droll Dry Earthy Effeminate Emotional Enigmatic Experimental Familial Folksy Formal Freewheeling Frugal Glamorous Guileless High-spirited Hurried Hypnotic Iconoclastic Idiosyncratic Impassive Impersonal Impressionable Intense Invisible Irreligious Irreverent Maternal Mellow Modern Moralistic Mystical Neutral Noncommittal Noncompetitive Obedient Old-fashioned Ordinary Outspoken Paternalistic Physical Placid Political Predictable Preoccupied Private Progressive Proud Pure Questioning Quiet Religious Reserved Restrained Retiring Sarcastic Self-conscious Sensual Skeptical Smooth Soft Solemn Solitary Stern Stoic Strict Stubborn Stylish Subjective Surprising Soft Tough Unaggressive Unambitious Unceremonious Unchanging Undemanding Unfathomable Unhurried Uninhibited Unpatriotic Unpredictable Unreligious Unsentimental Whimsical Abrasive Abrupt Agonizing Aimless Airy Aloof Amoral Angry Anxious Apathetic Arbitrary Argumentative Arrogant Artificial Asocial Assertive Astigmatic Barbaric Bewildered Bizarre Bland Blunt Boisterous Brittle Brutal Calculating Callous Cantankerous Careless Cautious Charmless Childish Clumsy Coarse Cold Colorless Complacent Complaintive Compulsive Conceited Condemnatory Conformist Confused Contemptible Conventional Cowardly Crafty Crass Crazy Criminal Critical Critical Cruel Cynical Decadent Deceitful Delicate Demanding Dependent Desperate Destructive Devious Difficult Dirty Disconcerting Discontented Discouraging Discourteous Dishonest Disloyal Disobedient Disorderly Disorganized Disputatious Disrespectful Disruptive Dissolute Dissonant Distractible Disturbing Dogmatic Domineering Dull Easily Discouraged Egocentric Enervated Envious Erratic Escapist Excitable Expedient Extravagant Extreme Faithless Fanatical Fanciful Fatalistic Fawning Fearful Fickle Fiery Fixed Flamboyant Foolish Forgetful Fraudulent Frightening Frivolous Gloomy Graceless Grand Greedy Grim Gullible Hateful Haughty Hedonistic Hesitant Hidebound High-handed Hostile Ignorant Imitative Impatient Impractical Imprudent Impulsive Inconsiderate Incurious Indecisive Indulgent Inert Inhibited Insecure In-sensitive Insincere Insulting Intolerant Irascible Irrational Irresponsible Irritable Lazy Libidinous Loquacious Malicious Mannerless Mawkish Mealy-mouthed Mechanical Meddlesome Melancholic Meretricious Messy Miserable Miserly Misguided Mistaken Money-minded Monstrous Moody Morbid Muddle-headed Naive Narcissistic Narrow Narrow-minded Natty Negativistic Neglectful Neurotic Nihilistic Obnoxious Obsessive Obvious Odd Offhand One-dimensional Opinionated Opportunistic Oppressed Outrageous Overimaginative Paranoid Passive Pedantic Perverse Petty Pharisaical Phlegmatic Plodding Pompous Possessive Power-hungry Predatory Prejudiced Presumptuous Pretentious Prim Procrastinating Profligate Provocative Pugnacious Puritanical Quirky Reactionary Reactive Regimental Regretful Repentant Repressed Ridiculous Rigid Ritualistic Rowdy Ruined Sadistic Sanctimonious Scheming Scornful Secretive Sedentary Selfish Self-indulgent Shallow Shortsighted Shy Silly Single-minded Sloppy Slow Sly Small-thinking Softheaded Sordid Steely Stiff Strong-willed Stupid Submissive Superficial Superstitious Suspicious Tactless Tasteless Tense Thievish Thoughtless Timid Transparent Treacherous Trendy Troublesome Unappreciative Uncaring Uncharitable Unconvincing Uncooperative Uncreative Uncreative Unfriendly Ungrateful Unhealthy Unimaginative Unimpressive Unlovable Unpolished Unprincipled Unrealistic Unreflective Unreliable Unrestrained Unself-critical Unstable Vacuous Vague Venal Venomous Vindictive Vulnerable Weak Weak-willed Well-

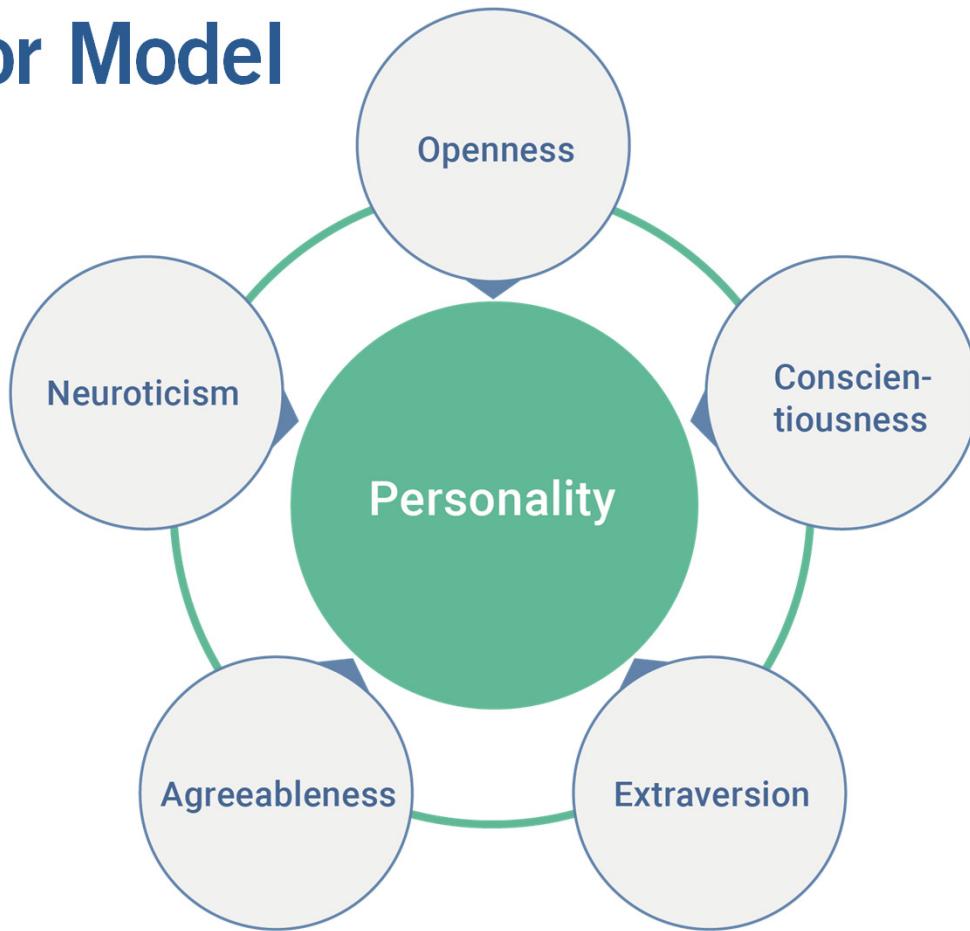
How Many Dimensions?



Eysenck's PEN Model of Personality

PSYCHOTICISM	EXTRAVERSION	Neuroticism
Aggressive	Sociable	Anxious
Assertive	Irresponsible	Depressed
Egocentric	Dominant	Guilt Feelings
Unsympathetic	Lack of reflection	Low self-esteem
Manipulative	Sensation-seeking	Tense
Achievement-oriented	Impulsive	Moody
Dogmatic	Risk-taking	Hypochondriac
Masculine	Expressive	Lack of autonomy
Tough-minded	Active	Obsessive

Five Factor Model



Psychometric Properties of the HEXACO Personality Inventory

Kiboom Lee
University of Calgary

Michael C. Ashton
Brock University

We introduce a personality inventory designed to measure six major dimensions of personality derived from lexical studies of personality structure. The HEXACO Personality Inventory (HEXACO-PI) consists of 24 facet-level personality trait scales that define the six personality factors named Honesty-Humility (H), Emotionality (E), Extraversion (X), Agreeableness (A), Conscientiousness (C), and Openness to Experience (O). In this validation study involving a sample of over 400 respondents, all HEXACO-PI scales showed high internal consistency reliabilities, conformed to the hypothesized six-factor structure, and showed adequate convergent validities with external variables. The HEXACO factor space, and the rotations of factors within that space, are discussed with reference to J. S. Wiggins' work on the circumplex.

HEXACO Personality Inventory

- The HEXACO model of personality conceptualizes human personality in terms of six dimensions.
 - Honesty-Humility (H)
 - Emotionality (E)
 - Extraversion (X)
 - Agreeableness (versus Anger) (A)
 - Conscientiousness (C)
 - Openness to Experience (O)
- It was proposed as alternative to the Big Five/FFM (Five Factor Model)

HEXACO Personality Inventory: 240 Questions

I love dangerous situations.
I need the approval of others.
I am the life of the party.
I am quick to judge others.
I make careless mistakes.
I seldom experience sudden intuitive insights.
I feel others' emotions.
I come up with something new.
I would not enjoy being a famous celebrity.
I tire out quickly.
I face danger confidently.
I react strongly to criticism.
I keep others at a distance.
I seek to derive less enjoyment from interacting with people than others do.
I prefer to eat at expensive restaurants.
I pretend to be concerned for others.
I often worry about things that turn out to be unimportant.
I would never go riding down a stretch of rapids in a canoe.
I rarely get irritated.
I demand quality.

I prefer to just let things happen.
I would not regret my behavior if I were to take advantage of someone impulsively.
I will not probe deeply into a subject.
I am sensitive to the needs of others.
I say little.
I don't know much about history.
I suspect that my facial expressions give me away when I feel sad.
I am good at making impromptu speeches.
I pay too little attention to details.
I do things without thinking of the consequences.
I maintain high energy throughout the day.
I have an eye for detail.
I have excellent ideas.
I am usually a patient person.

I steal things.
I need reassurance.
I boast about my virtues.
I do not like art.
I don't think that I'm better than other people.
I feel comfortable around people.
I seldom get mad.
I get upset easily.
I talk to a lot of different people at parties.
I when interacting with a group of people, am often bothered by at least one of them.
I try to forgive and forget.
I make a fool of myself.
I know that my ideas sometimes surprise me.
I get...
I rare...
I have...
I en...
I mak...
I wan...
I lov...

I rarely feel angry with people.
I have a strong personality.
I have a good word for everyone.
I tell people about it when I'm irritated.
I have great stamina.
I love luxury.
I would feel very badly for a long time if I were to steal from someone.
I don't finish the things that I start.
I find it difficult to approach others.
I usually like to spend my free time with people.
I admire a really clever scam.
I distrust people.
I can't do without the company of others.
I seldom get emotional.
I enjoy intellectual games.
I am mainly interested in money.

I find it hard to forgive others.
I don't know why I do some of the things I do.
I like to attract attention.
I have a vivid imagination.
I see myself as an average person.
I rarely cry during sad movies.
I work hard.
I don't worry about things that have already happened.
I wish to stay young forever.
I am hard to reason with.
I love to think up new ways of doing things.
I speak softly.
I do not have a good imagination.
I feel healthy and vibrant most of the

time.
I feel angry at someone.
I let people push me around to help them feel important.
I would love to explore strange places.
I have read the great literary classics.
I want to be liked.
I don't talk a lot.
I seldom feel weepy while reading the sad part of a story.
I stop when work becomes too difficult.
I smile a lot.
I would be afraid to give a speech in public.
I seek status.
I am deeply moved by others' misfortunes.
I am hard to convince.
I push myself very hard to succeed.
I switch my loyalties when I feel like it.

I adjust easily.
I am willing to take risks.
I believe in the importance of art.
I rarely feel depressed.
I don't strive for elegance in my appearance.
I swim against the current.
I rebel against authority.
I get angry easily.
I get upset by unpleasant thoughts that come into my mind.
I return extra change when a cashier makes a mistake.
I get deeply immersed in music.
I bottle up my feelings.
I do not enjoy watching dance performances.

I might not notice.
I would hate to be considered odd or strange.
I criticize others' shortcomings.
I am good at taking advice.
I am usually active and full of energy.
I would not enjoy a job that involves a lot of social interaction.
I often forget to put things back in their proper place.
I find it necessary to please the people who have power.
I have little to say.
I feel that I have a lot of inner strength.
I hang around doing nothing.
I am a physical coward.

I rarely enjoy being with people.
I play a role in order to impress people.
I try to follow the rules.
I don't mind being the center of attention.
I get started quickly on doing a job.
I consider myself an average person.
I get upset if others change the way that I have arranged things.
I am inclined to forgive others.
I speak ill of others.
I am likely to show off if I get the chance.
I enjoy being thought of as a normal mainstream person.
I am not easily disturbed by events.

Others will
I quickly lose interest in the tasks I start.
I complete tasks successfully.
I am interested in science.
I like order.
I do things according to a plan.
I show my sadness.
I tell other people what they want to hear so that they will do what I want them to do.
I lose my temper.
I act impulsively when something is bothering me.
I don't bother worrying about political and social problems.
I am easily annoyed.
I do not like concerts.
I don't pretend to be more than I am.
I am full of ideas.

240 Questions → 6 Dimensions

HEXACO Personality Inventory: 240 Questions

All variables are recorded on a seven-point scale

- 1 = strongly disagree
- 2 = disagree
- 3 = slightly disagree
- 4 = neutral
- 5 = slightly agree
- 6 = agree
- 7 = strongly agree



This website provides a collection of interactive personality tests with detailed results that can be taken for personal entertainment or to learn more about personality assessment. These tests range from very serious and widely used scientific instruments popular psychology to self produced quizzes. A special focus is given to the strengths, weaknesses and validity of the various systems.

Recommended test for scientific validity

Big Five Personality Test: The general consensus in academic psychology is that there are five fundamental personality traits. This model is assumed in most personality research, and is the basis of many of the most well regarded tests employed by psychologists who maintain close connections with academia. The "big five" tend to not be popular in consumer focused personality assessment or self-help because to many people the feedback of the model seems relatively basic. This test uses public domain scales from the International Personality Item Pool.

Recommended test for personal enjoyment

Open Extended Jungian Type Scales: The system of personality types proposed by Carl Jung (1921) and later refined by C. Myers and I. M. Briggs has become an extremely widely used personality theory in self-help, business management, counselling and spiritual development contexts, but it is not commonly used in academic research where, like all type theories, it is treated sceptically. The system produces 16 personality types on the basis of four dichotomies and is the system used in the Myers Briggs Type Indicator and Keirsey Temperament Sorter instruments, among many others. The OEJS is a free and open source measure of the four dichotomies which yields an equivalent result to the usual tests.

Today's Objectives

“Exploratory Factor Discovery”

- Discovery of **latent** (hidden) concepts.
- Easy-to-interpret, meaningful factors
- Homogenous clusters
- Stable dimensions
- “Careful” dimension reduction
- Computationally tractable

lateo (present infinitive latere, perfect active latui); second conjugation, no passive

- I am concealed or in hiding, lurk, skulk.
 - Latet anguis in herba. – A snake hides in the grass.
 - Sub nomine pacis bellum latet. – War lurks under the name of peace.
- I am hidden and in safety.
- I keep out of sight.
- I live in concealment; live retired.
- I escape notice, remain unknown.
 - Bene qui latuit, bene vixit. – He who has well remained unknown has lived well.
- I am obscure or unknown, lie hidden.
 - Id qua ratione consecutus sit latet. – It is unknown how he obtained that.

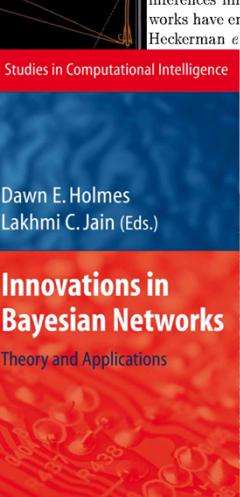
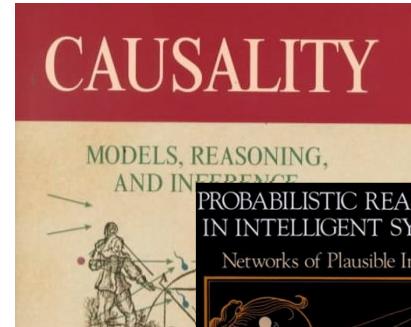
Methodology?





Why Bayesian Networks & BayesiaLab?

The New Paradigm: Bayesian Networks



BAYESIAN NETWORKS*

Judea Pearl

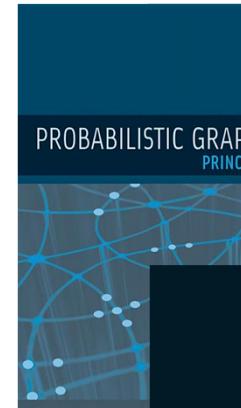
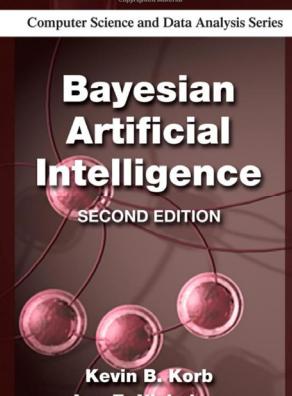
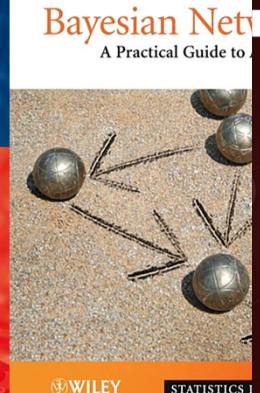
Cognitive Systems Laboratory

Computer Science Department

University of California, Los Angeles, CA 90024

judea@cs.ucla.edu

Bayesian networks were developed in the late 1970's to model distributed processing in reading comprehension, where both semantical expectancies and causal influences could be combined to form a coherent interpretation. They have since become a widely used tool for modeling causal relationships in many domains. This book provides a comprehensive introduction to the theory and application of Bayesian networks. It covers the basic concepts, algorithms, and applications of Bayesian networks, and provides a detailed treatment of advanced topics such as causal reasoning, prediction, and learning.

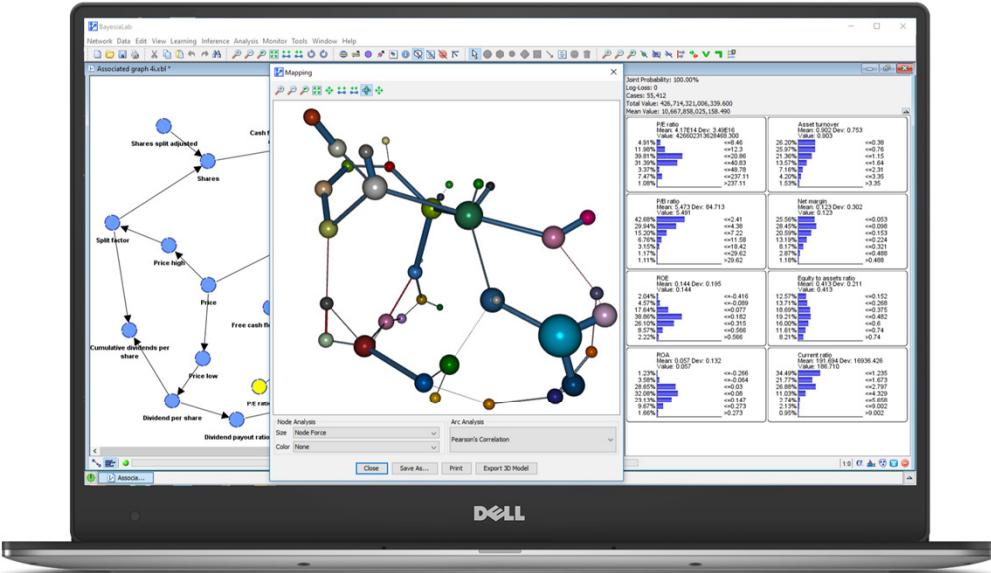


The top right corner shows the BayesiaLab logo and the text 'BAYESIA'. Below it is a screenshot of a laptop displaying the BayesiaLab software interface, which shows a network of nodes and connections. To the right of the laptop, the text 'Bayesian Networks & BayesiaLab' and 'STEFAN CONRADY | LIONEL JOUFFE' are visible. At the bottom, there is a link 'A Practical Introduction for Researchers' and the URLs 'bayesia.us • bayesia.com • bayesia.sg'.

Mathematical Formalism → Research Software



BAYESIALAB

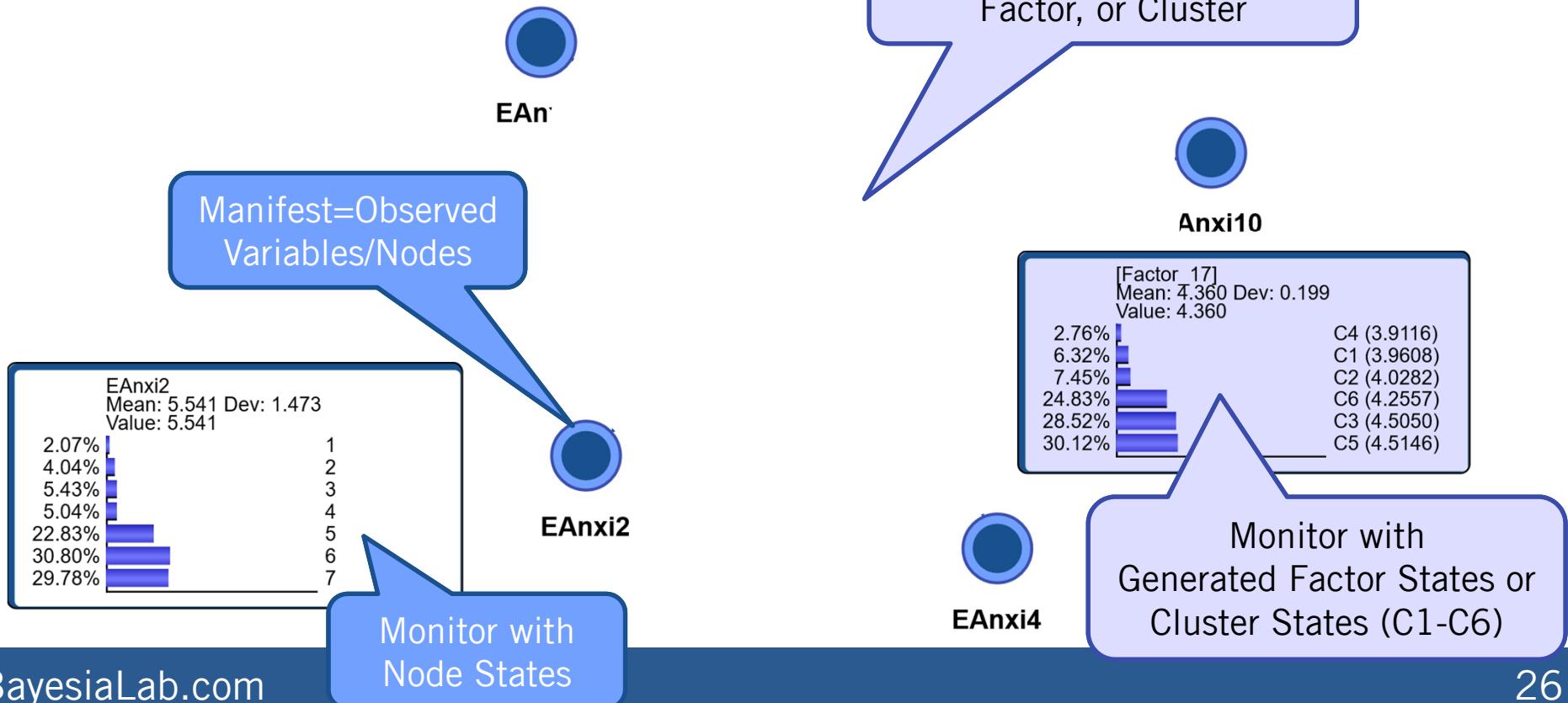


A desktop software for:

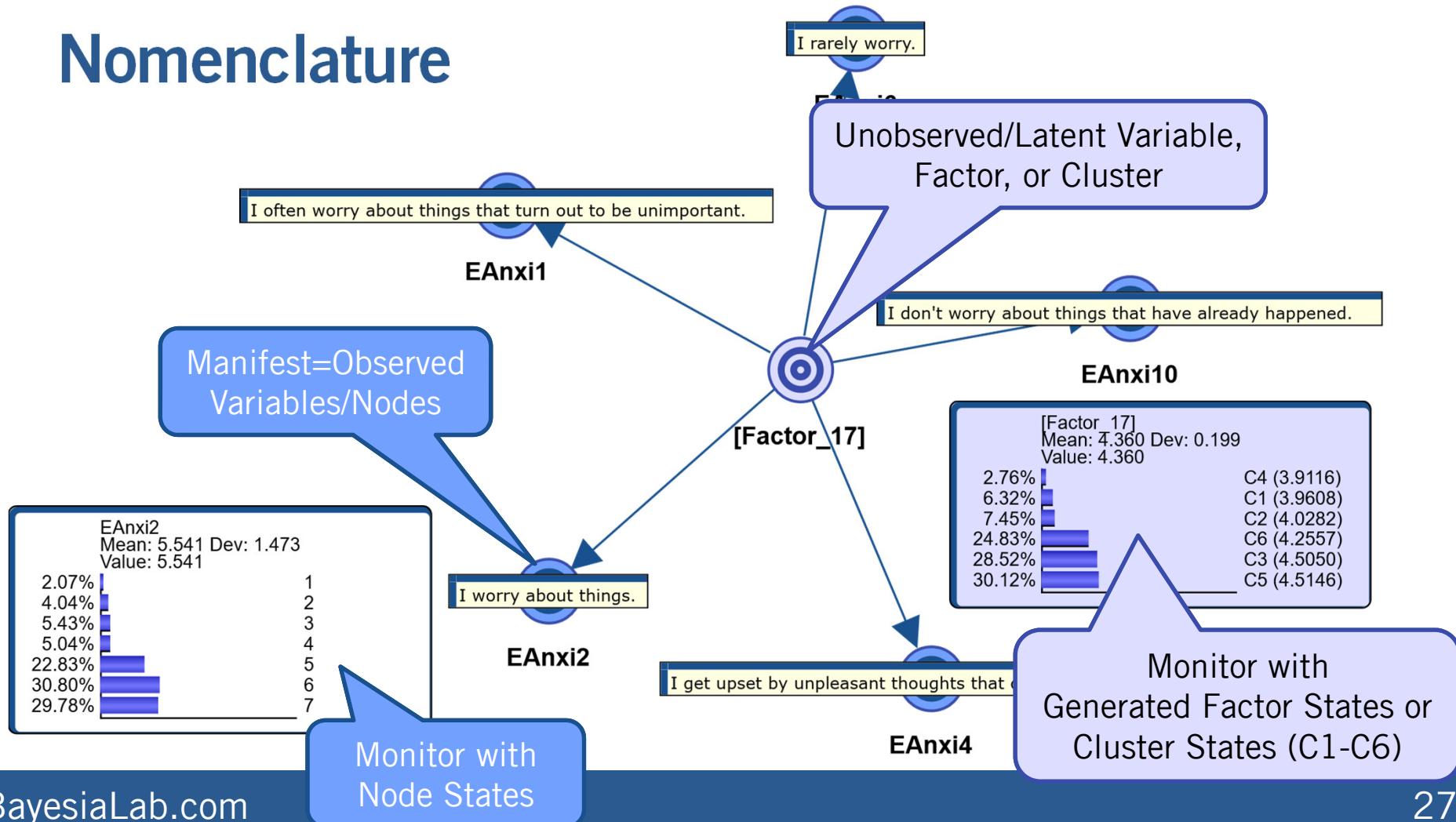
- encoding
- learning
- editing
- performing inference
- analyzing
- simulating
- optimizing

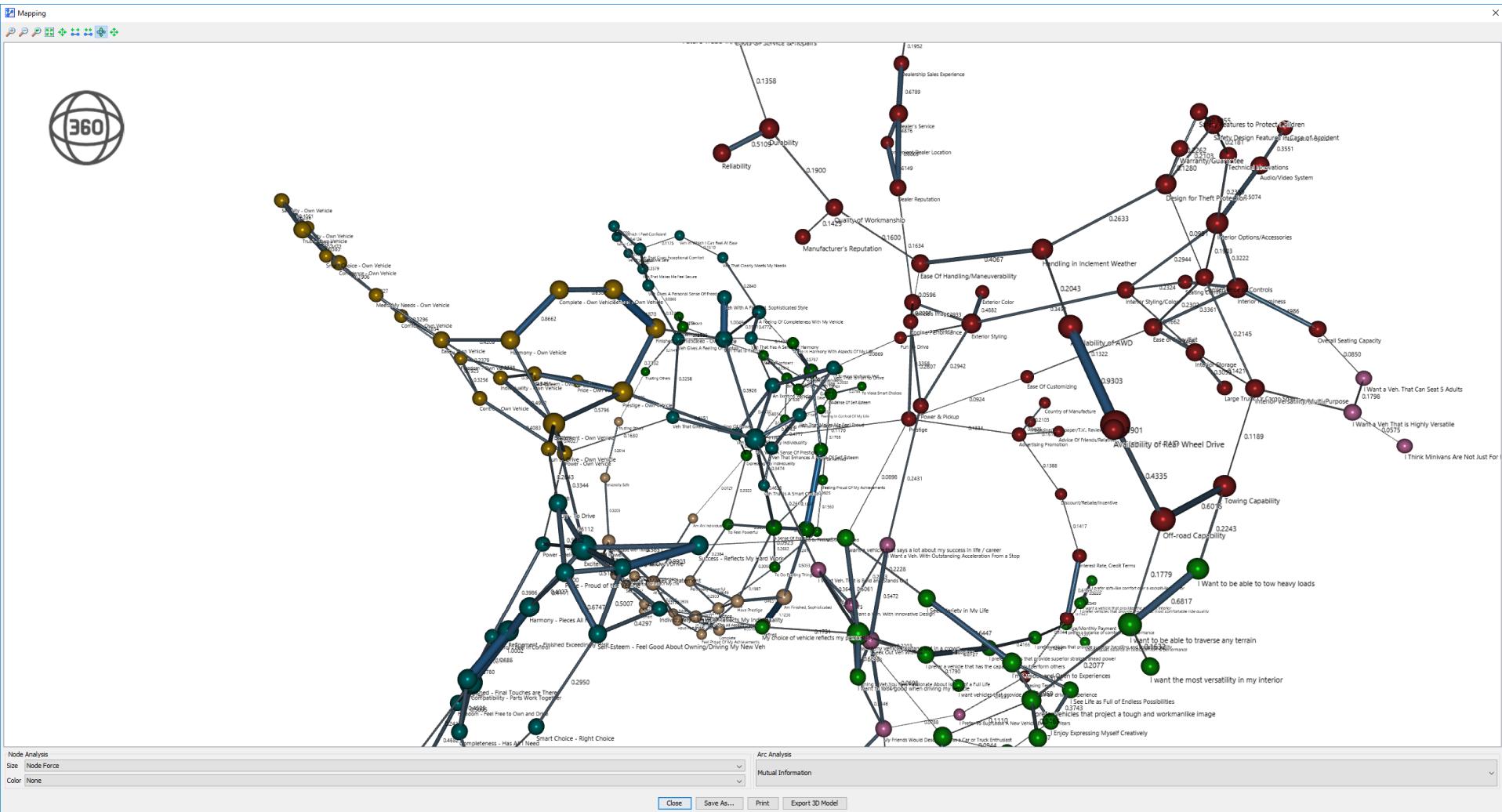
with Bayesian networks.

Nomenclature



Nomenclature





Bayesian Networks

Key Concepts Important for Factors Analysis

- As a Bayesian network represents a joint probability distribution, we can easily compute several information-theoretic measures with BayesiaLab:
 - Entropy
 - Mutual Information
 - **Arc Force**
 - Node Force
 - Etc.

**MAY THE
ARC FORCE
BE WTH YOU!**

Bayesian Networks

Key Concepts Important for Factors Analysis

- As a Bayesian network represents a joint probability distribution, we can easily compute several information-theoretic measures:

- Entropy
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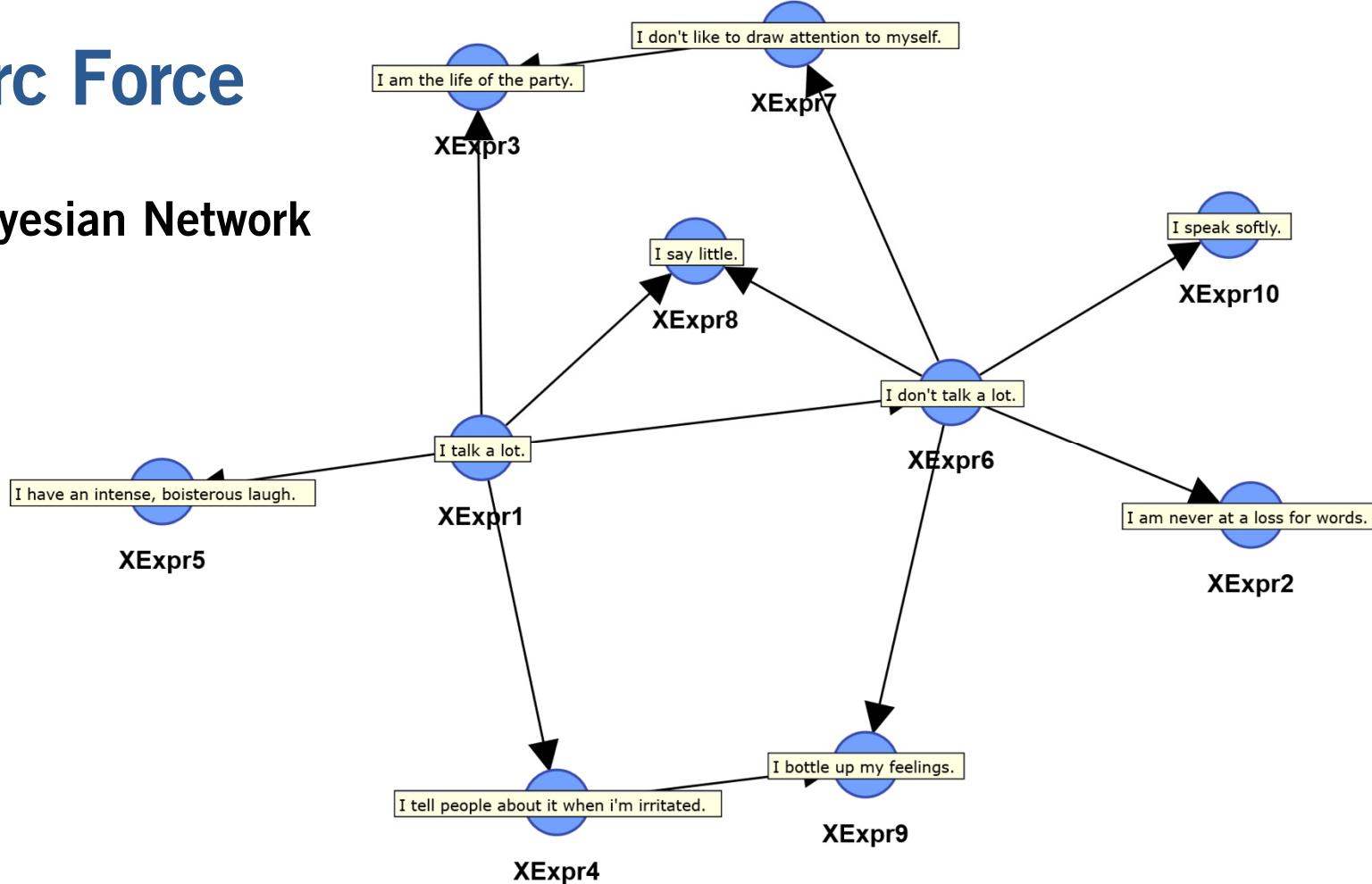


Arc Force, a Measure of “Arc Importance”

- Arc Force is more formally known as “Kullback–Leibler Divergence.”
- It is the difference or distance in the joint distributions Q and P denoted $D_{KL}(P \parallel Q)$

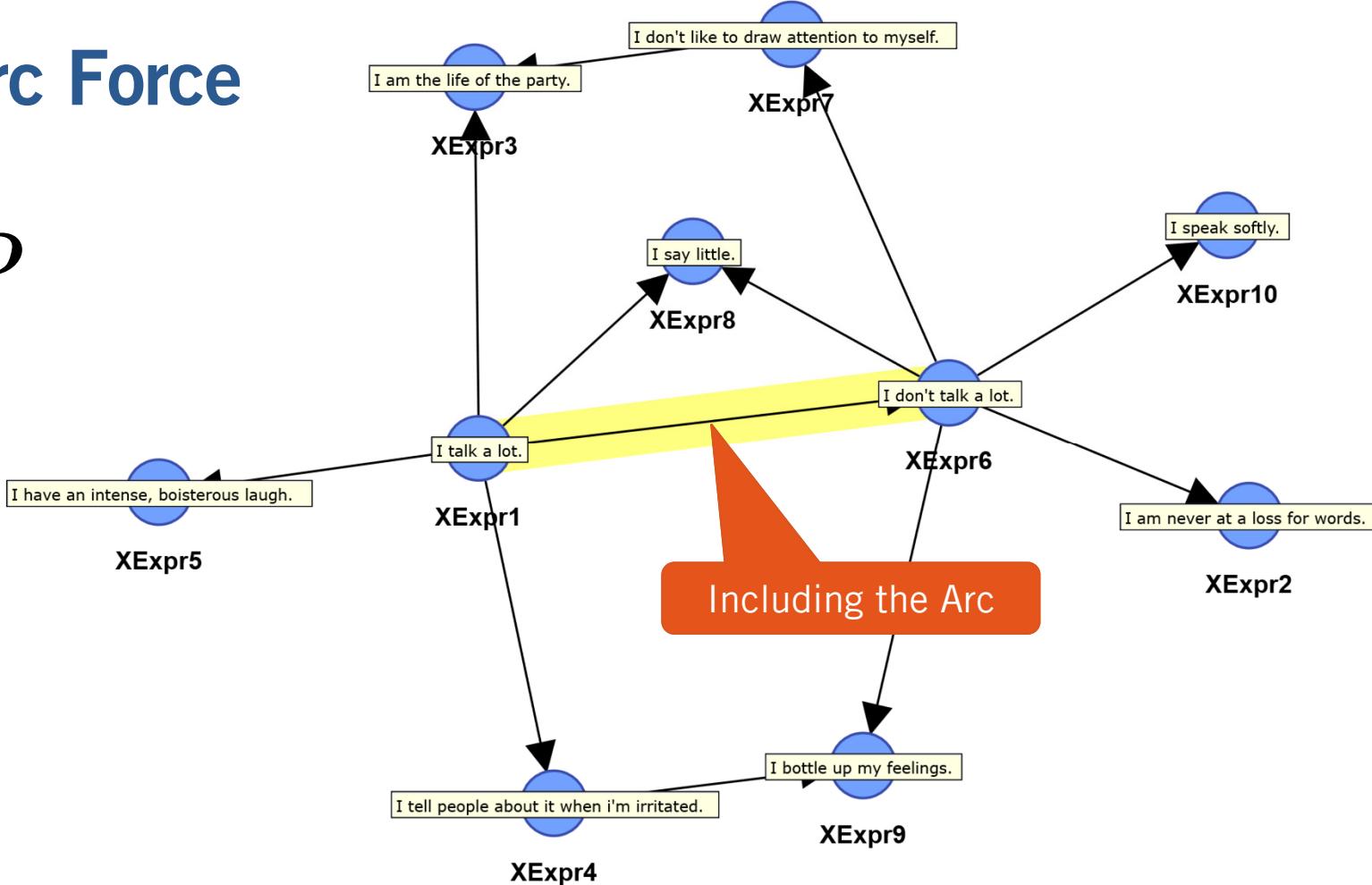
Arc Force

Bayesian Network



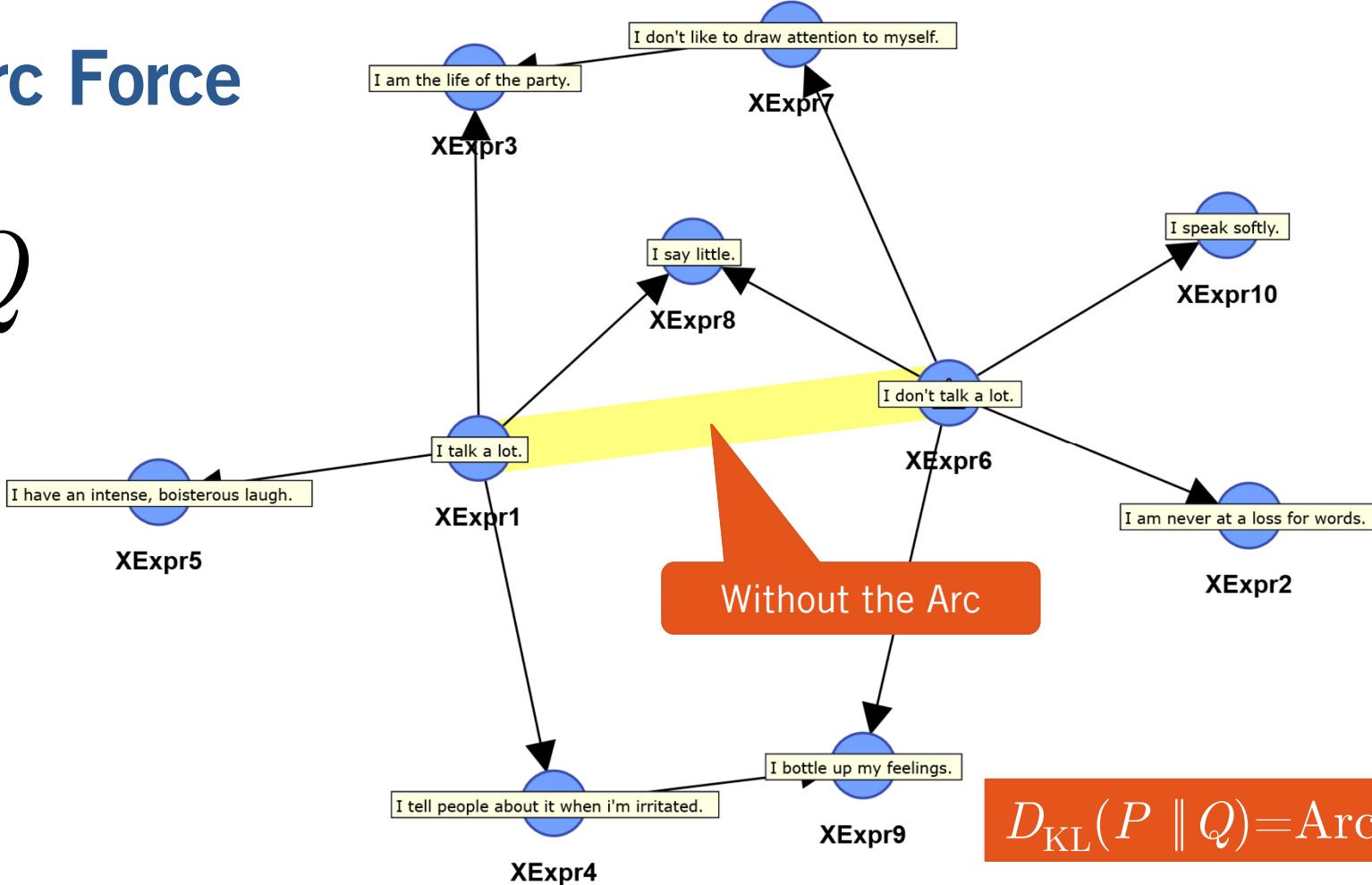
Arc Force

P



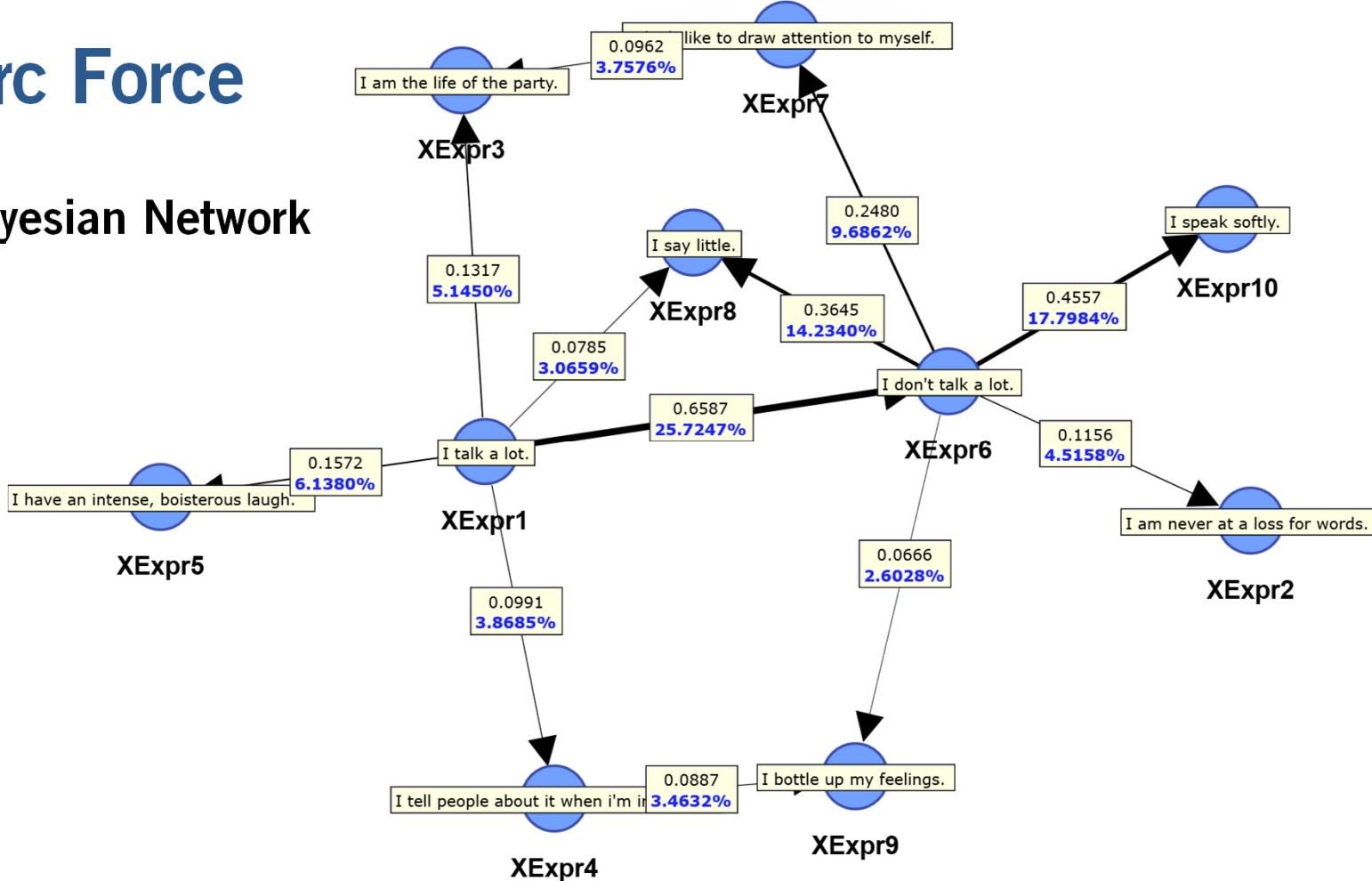
Arc Force

Q



Arc Force

Bayesian Network



Arc Force

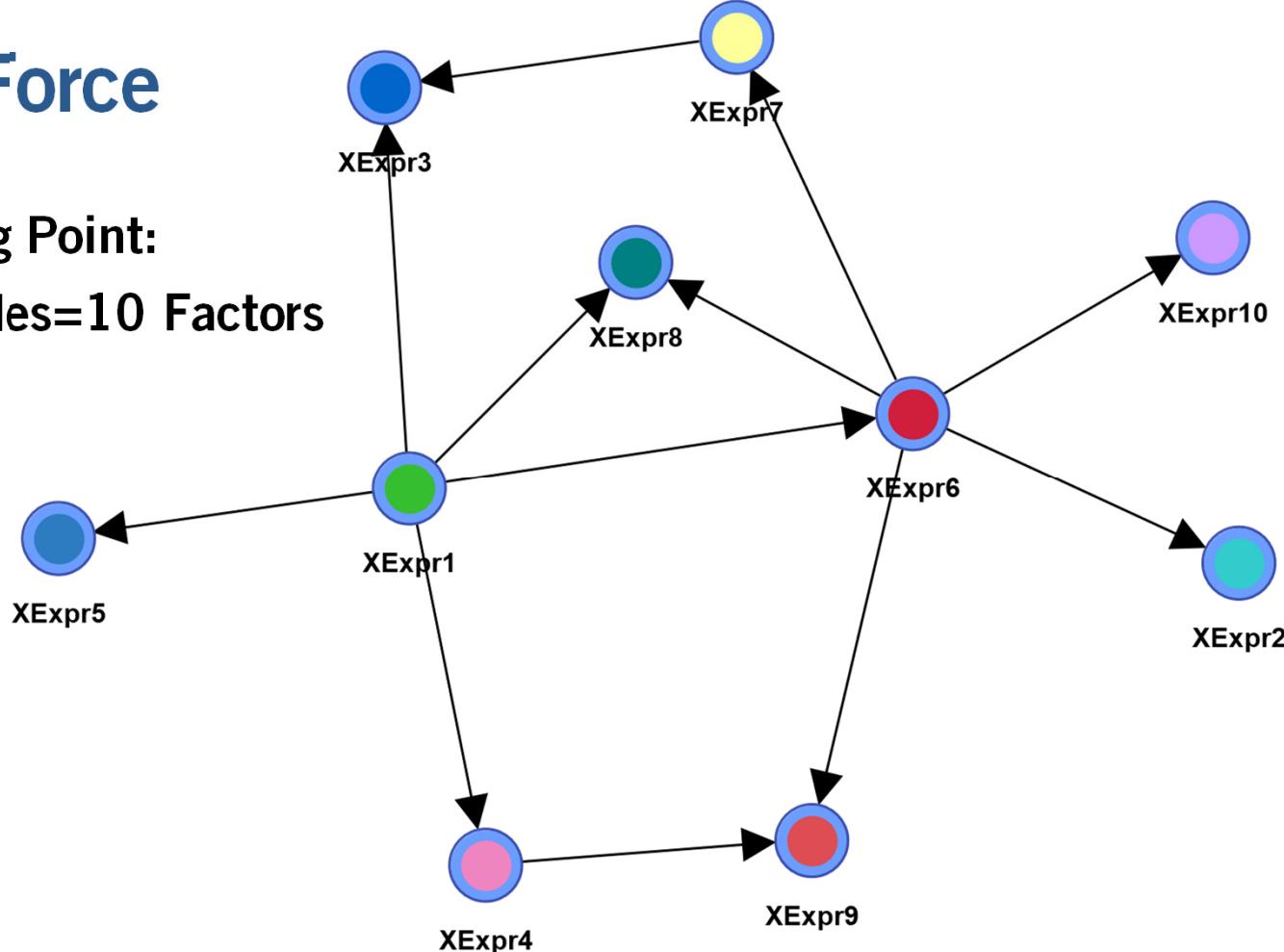


How does this
help with
clustering?

Arc Force

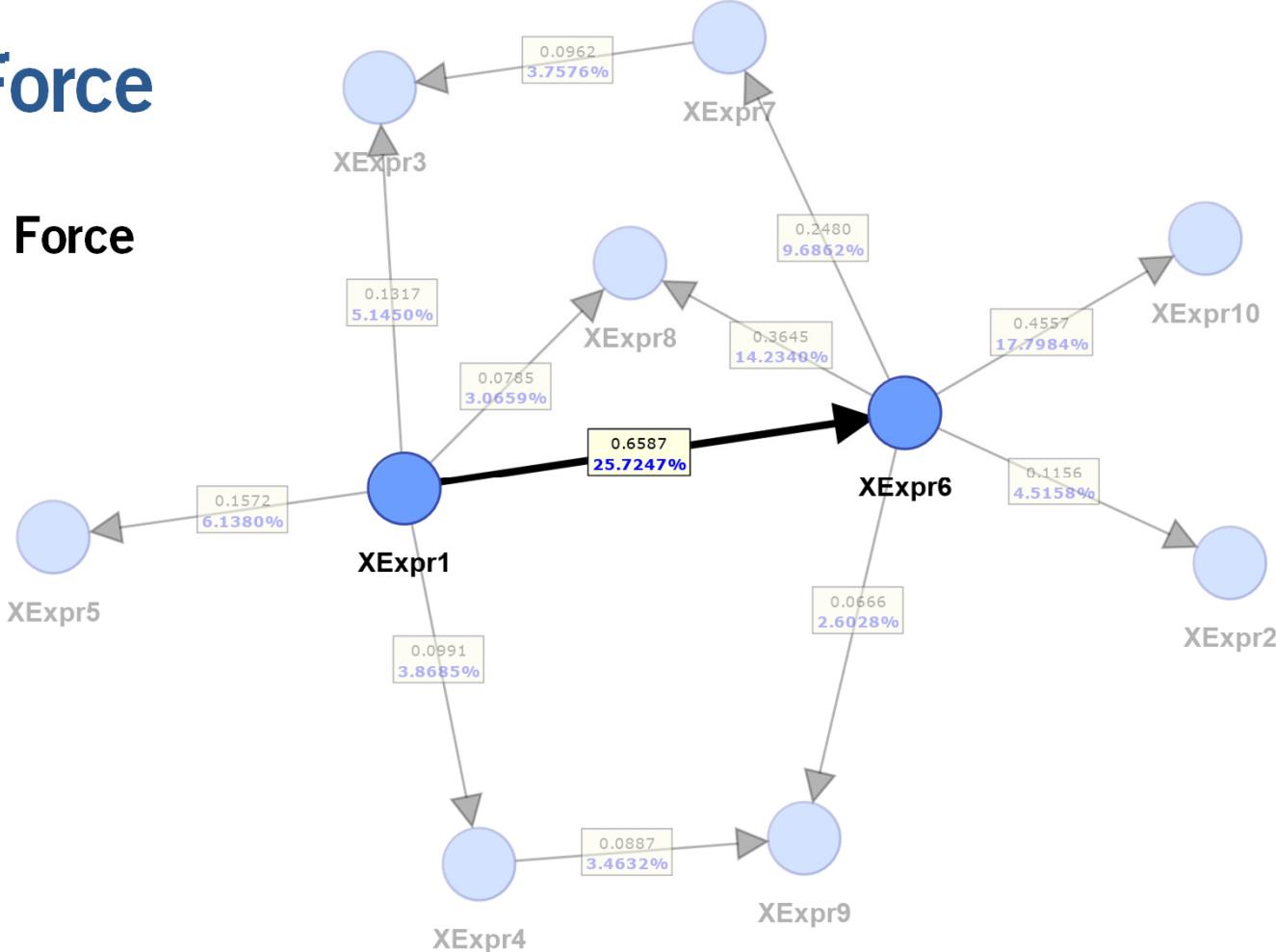
Starting Point:

10 Nodes=10 Factors



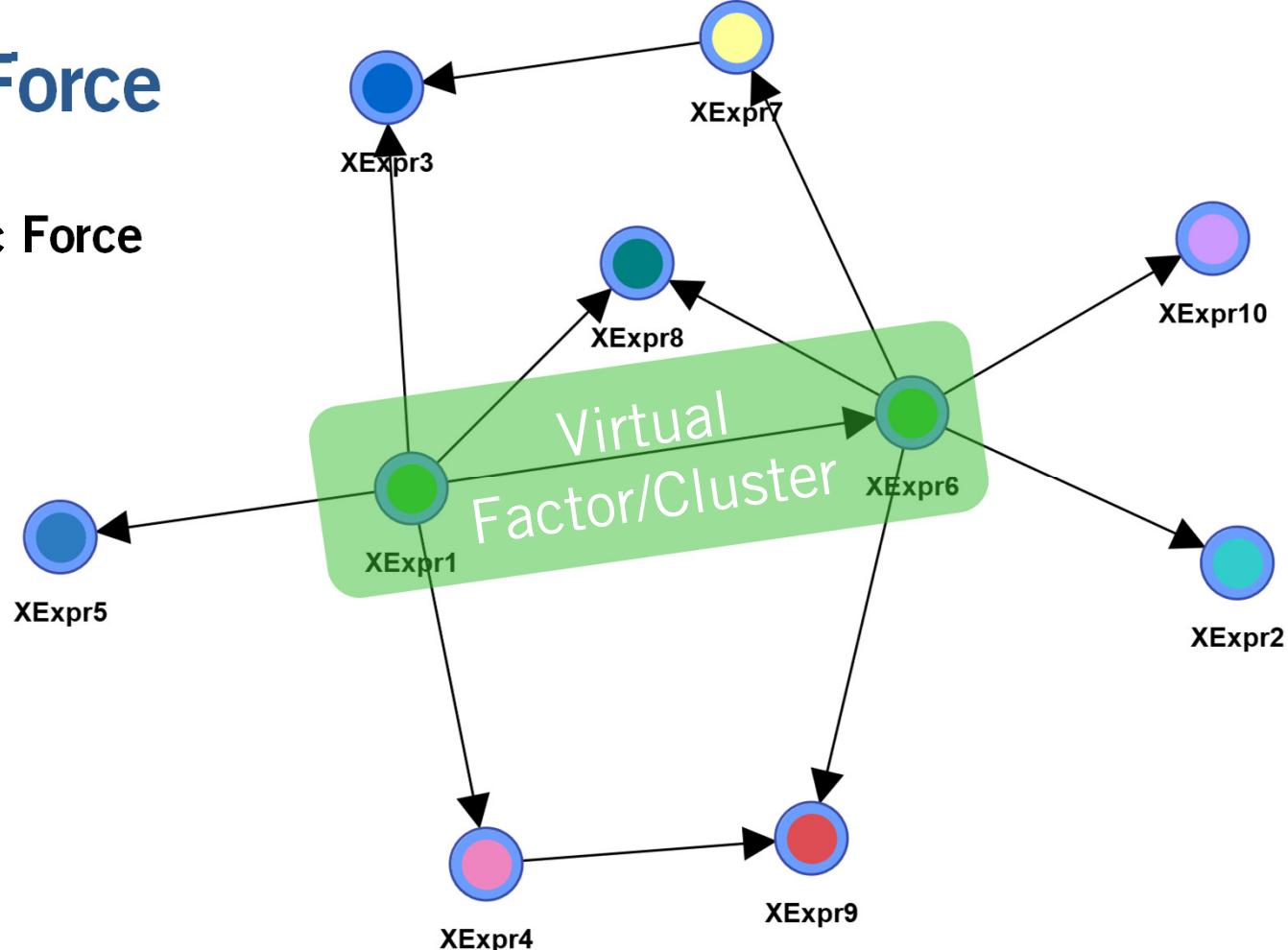
Arc Force

Top Arc Force



Arc Force

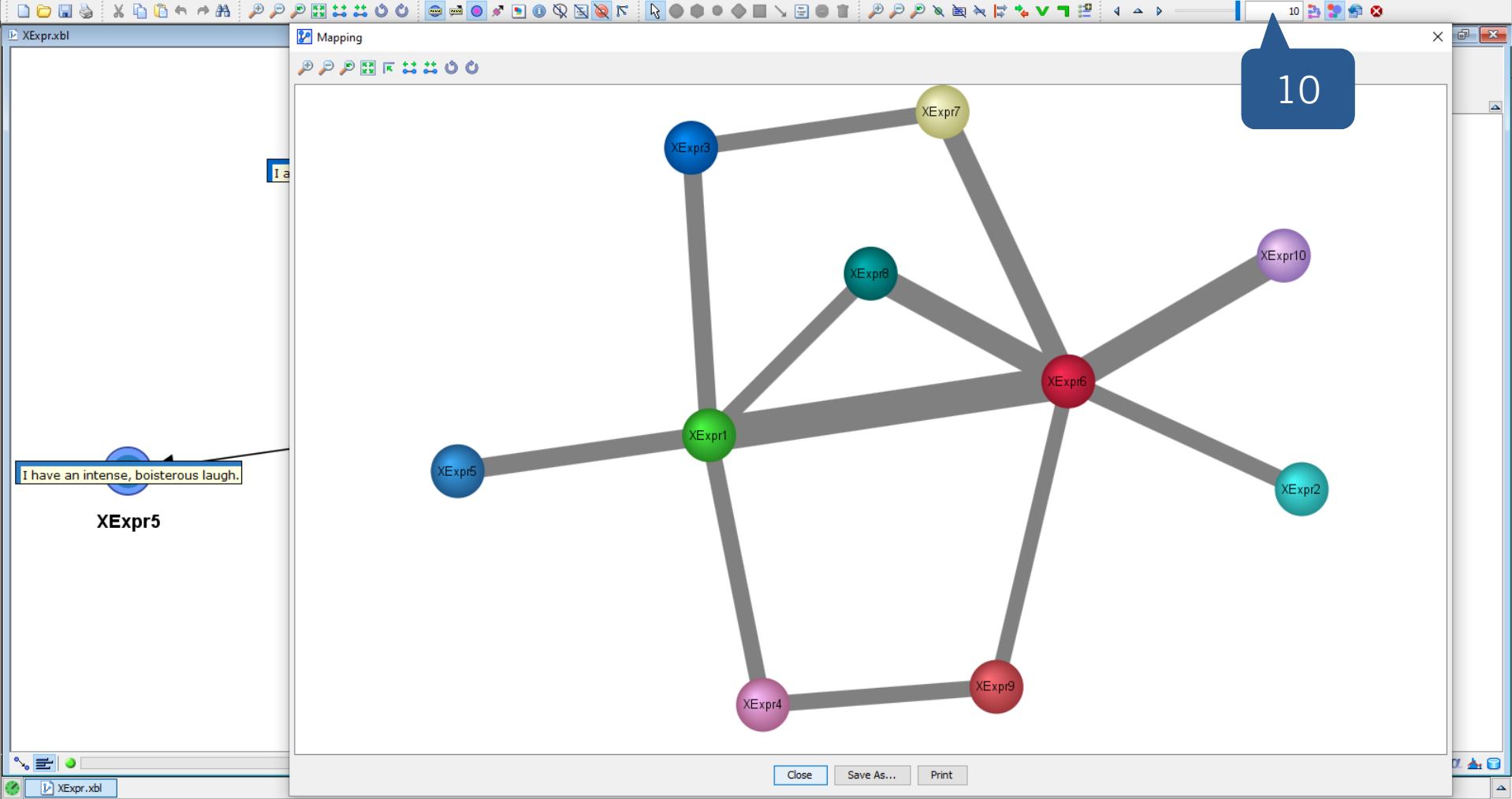
Top Arc Force

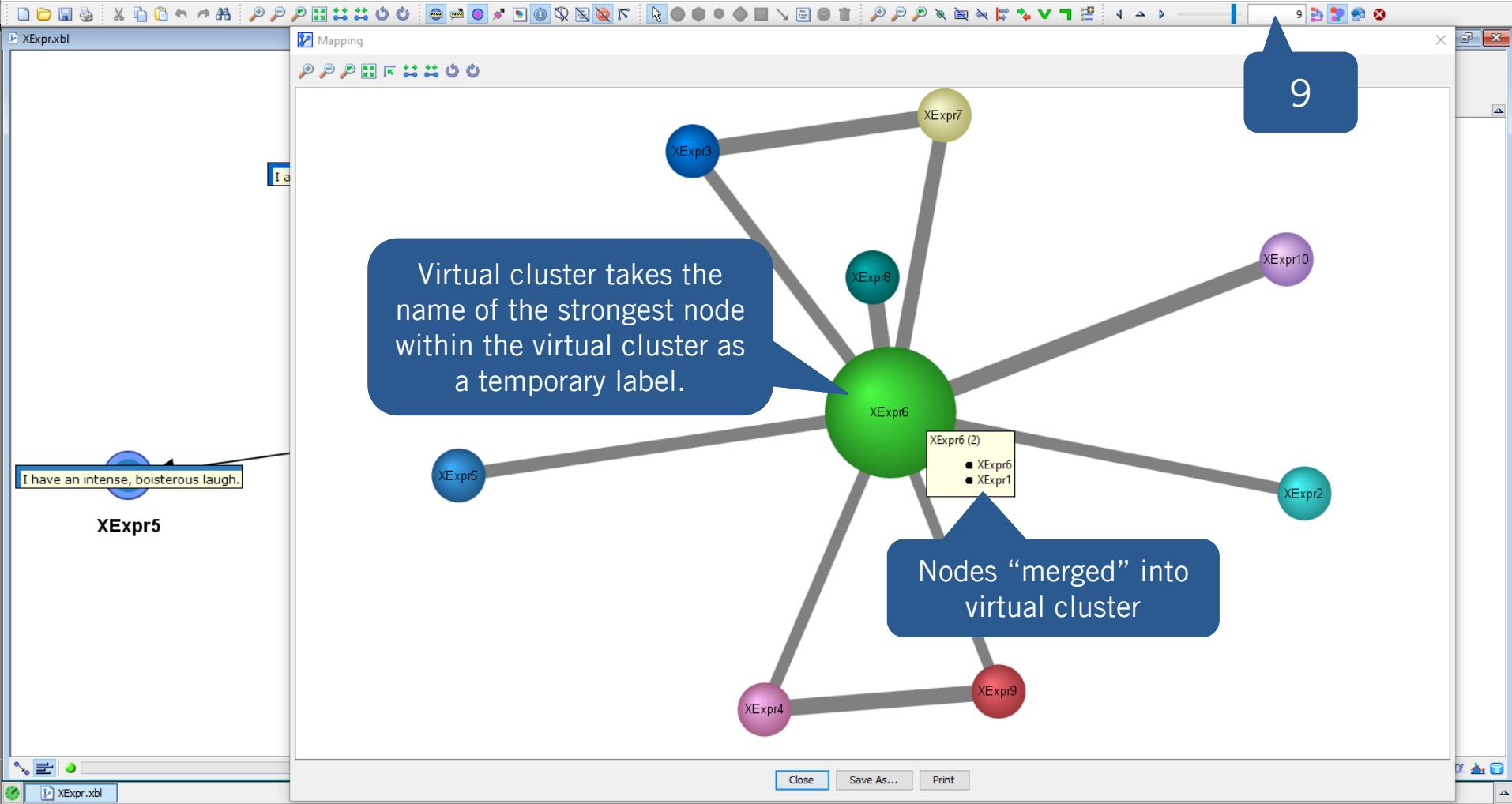


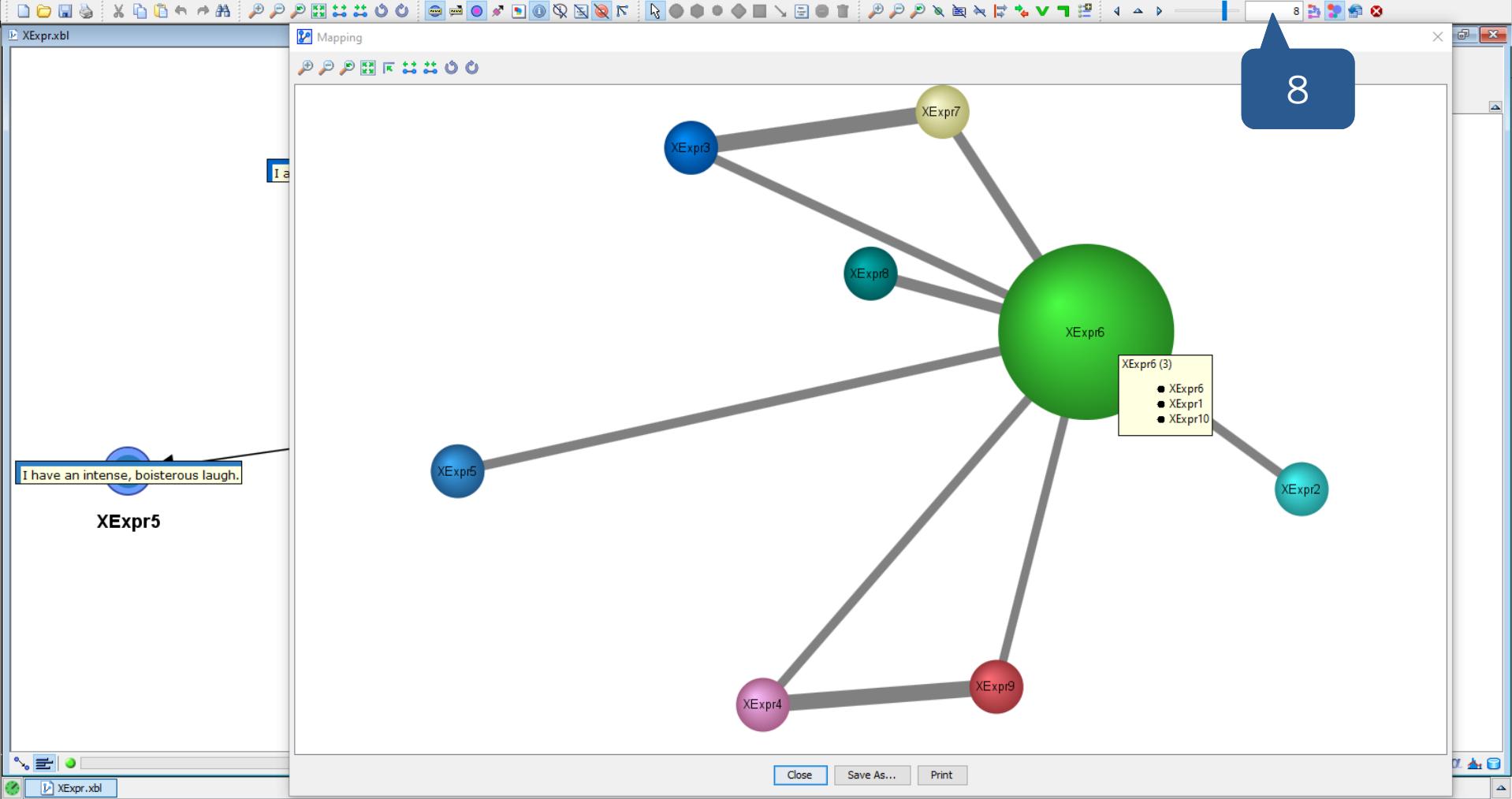
Arc Force

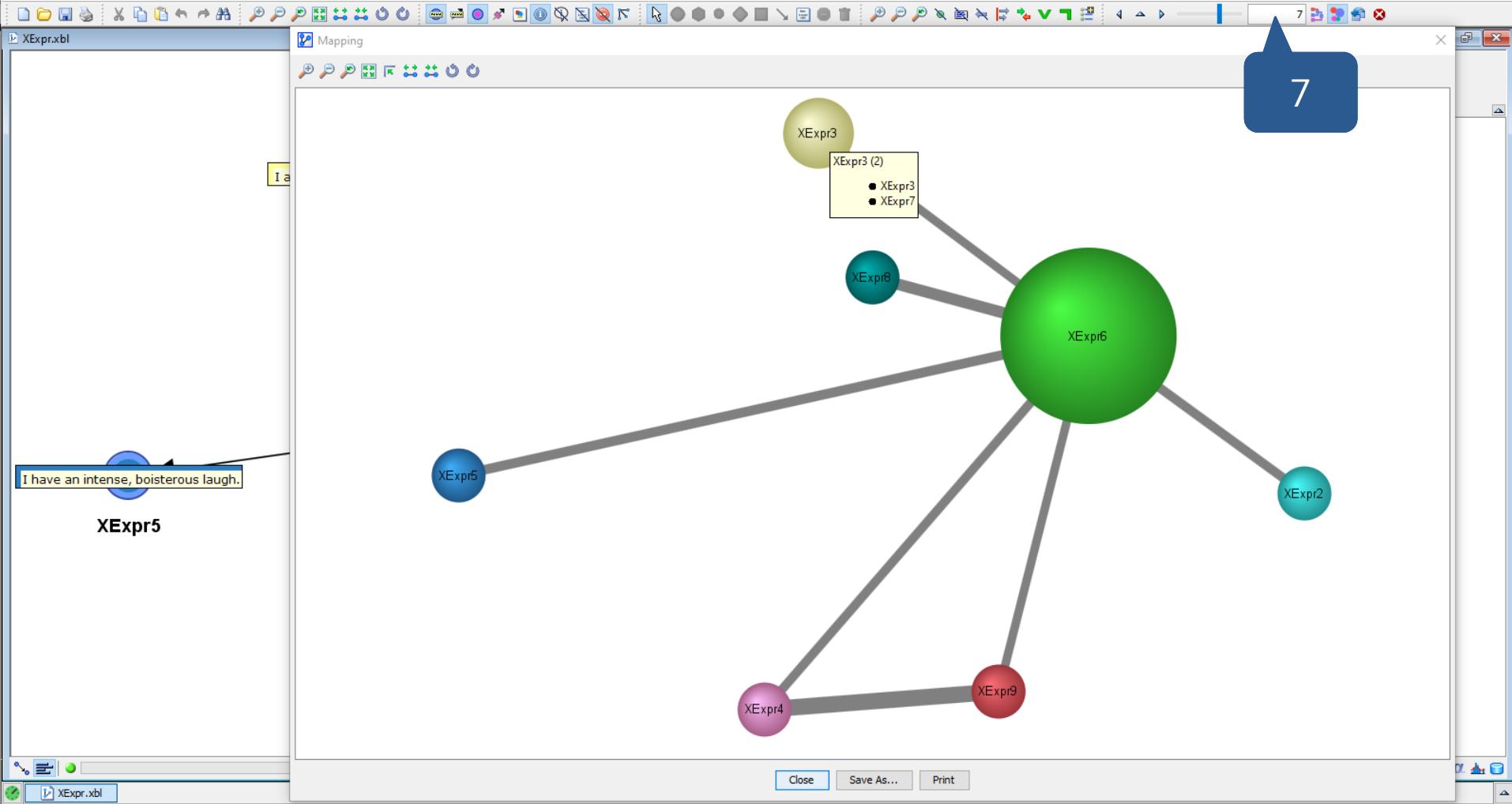
Using Arc Force for Variable Clustering

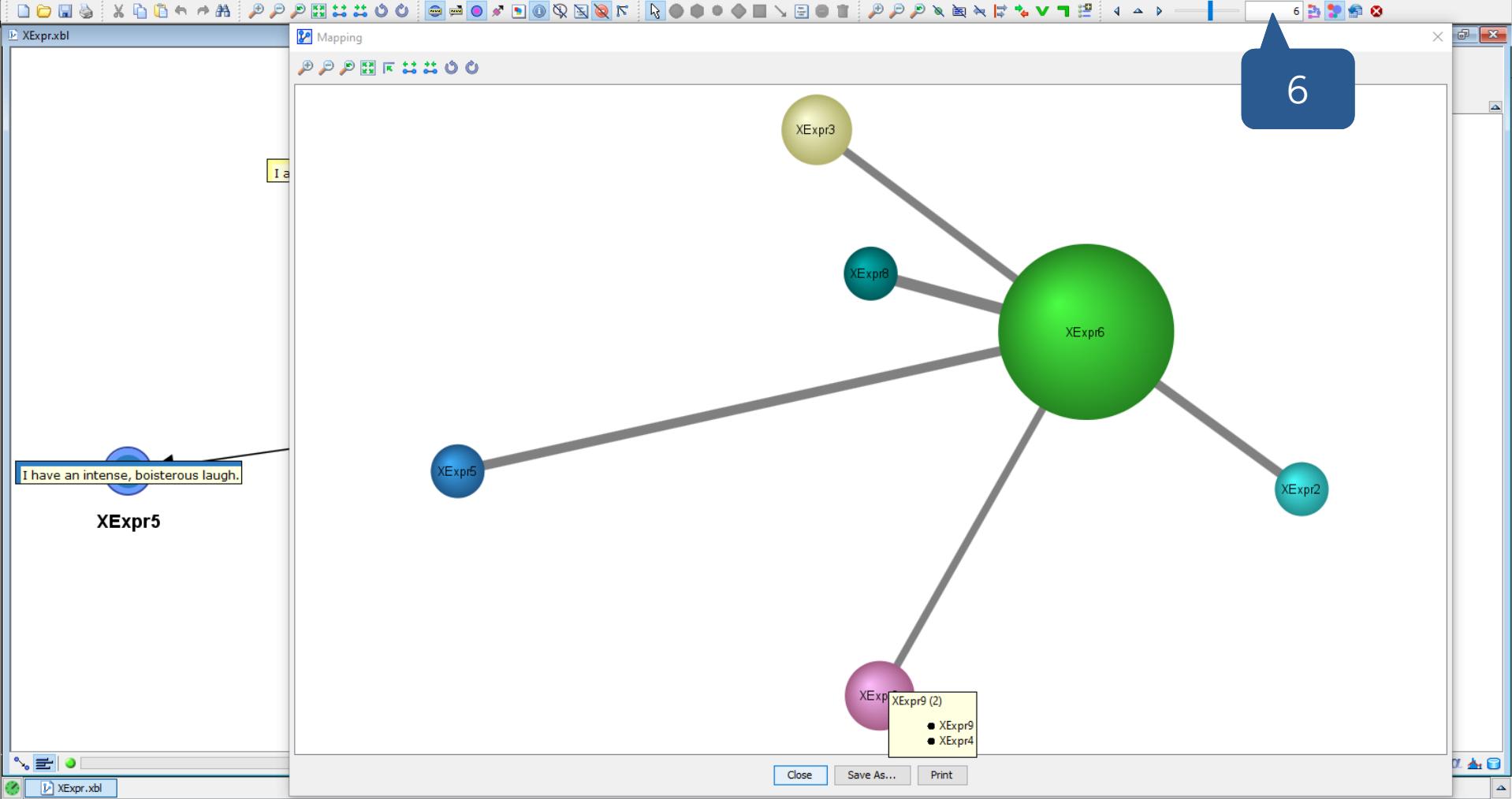
- BayesiaLab's Variable Clustering is a hierarchical agglomerative clustering algorithm that uses Arc Force (i.e., the Kullback-Leibler Divergence) for computing the distance between nodes.
- At the start of Variable Clustering, each manifest variable is treated as a distinct Factor or Cluster.
- The clustering algorithm proceeds iteratively by merging the “closest” Factors/Clusters into a new Factor/Cluster.

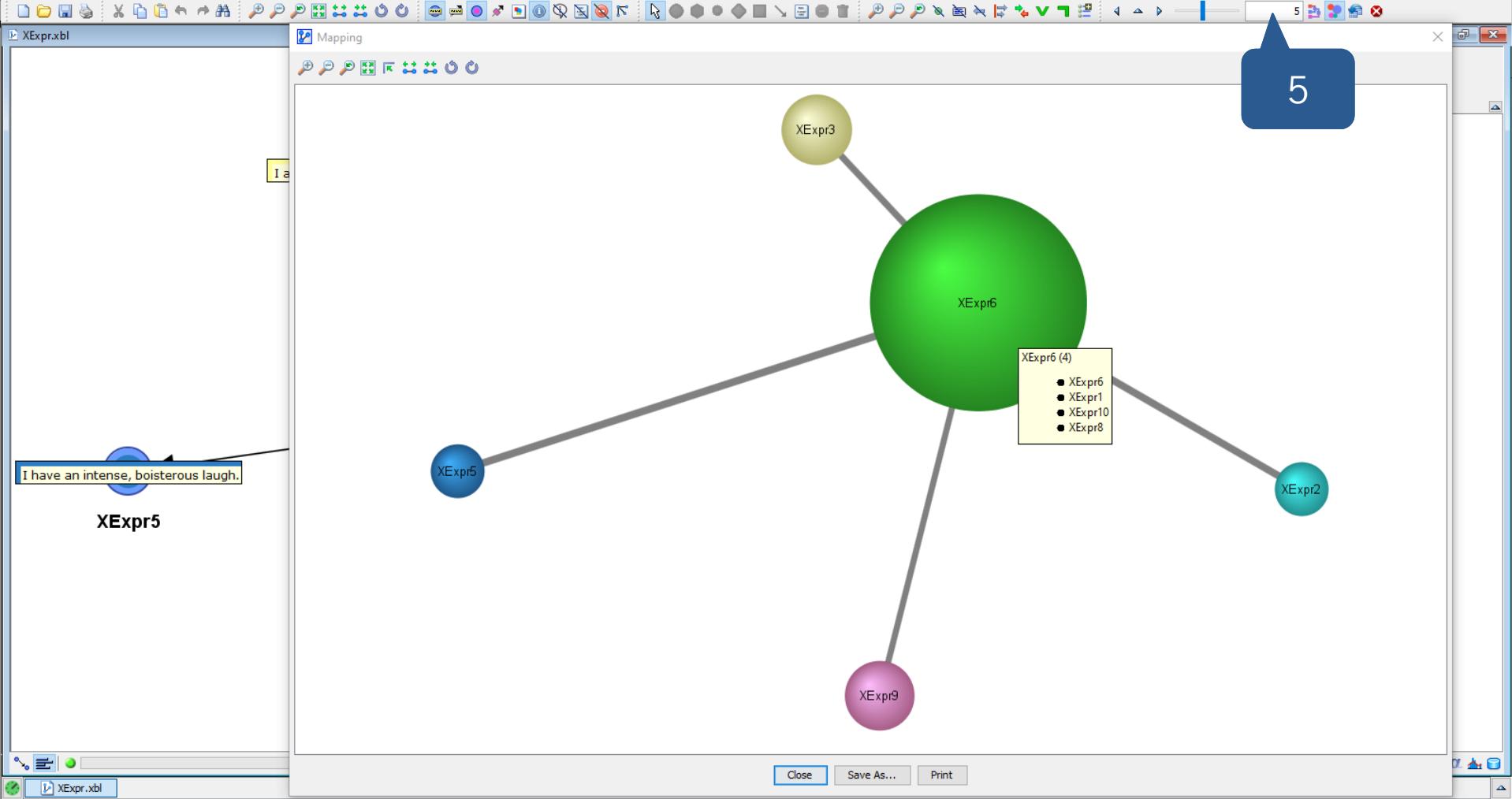


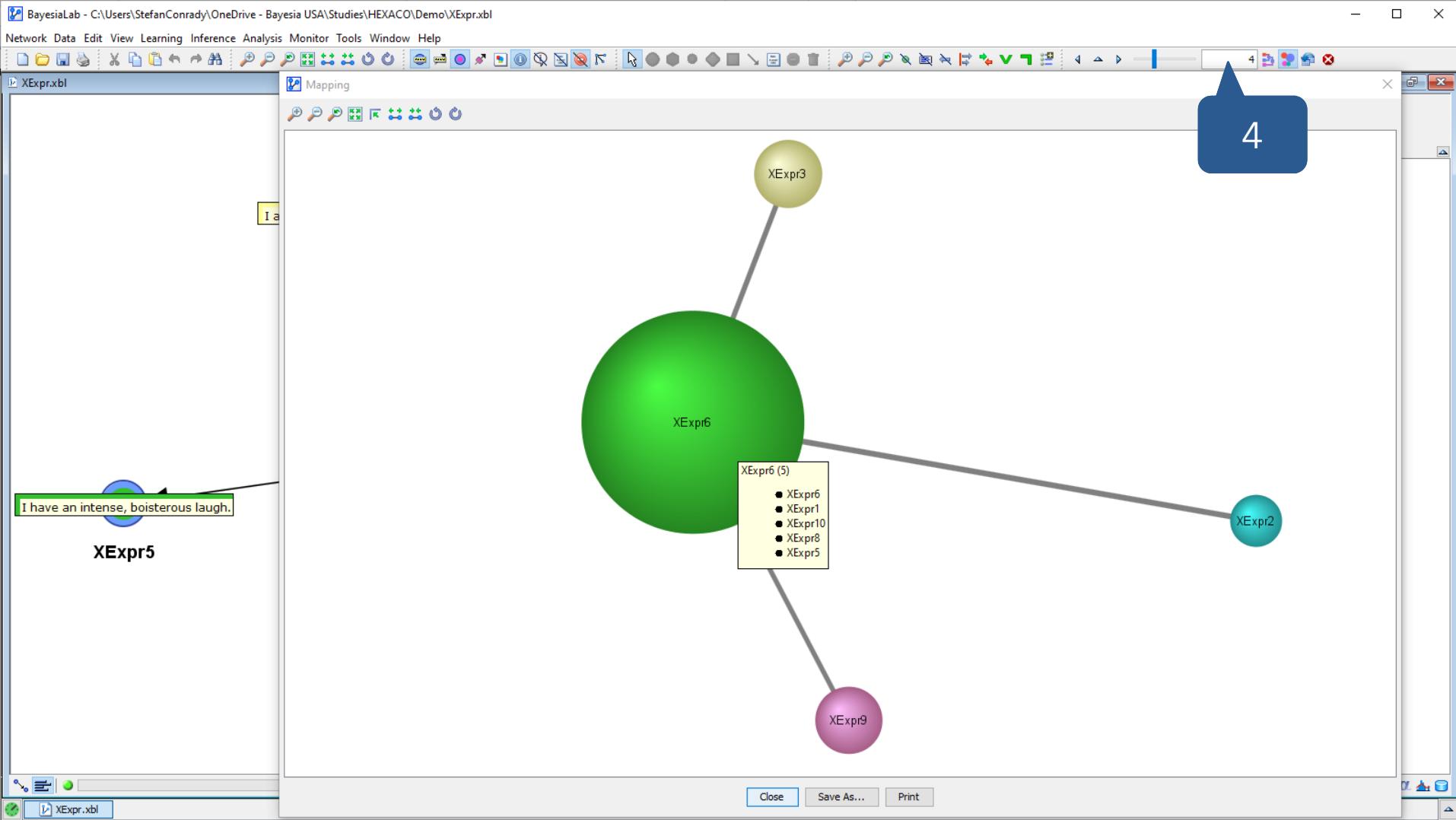


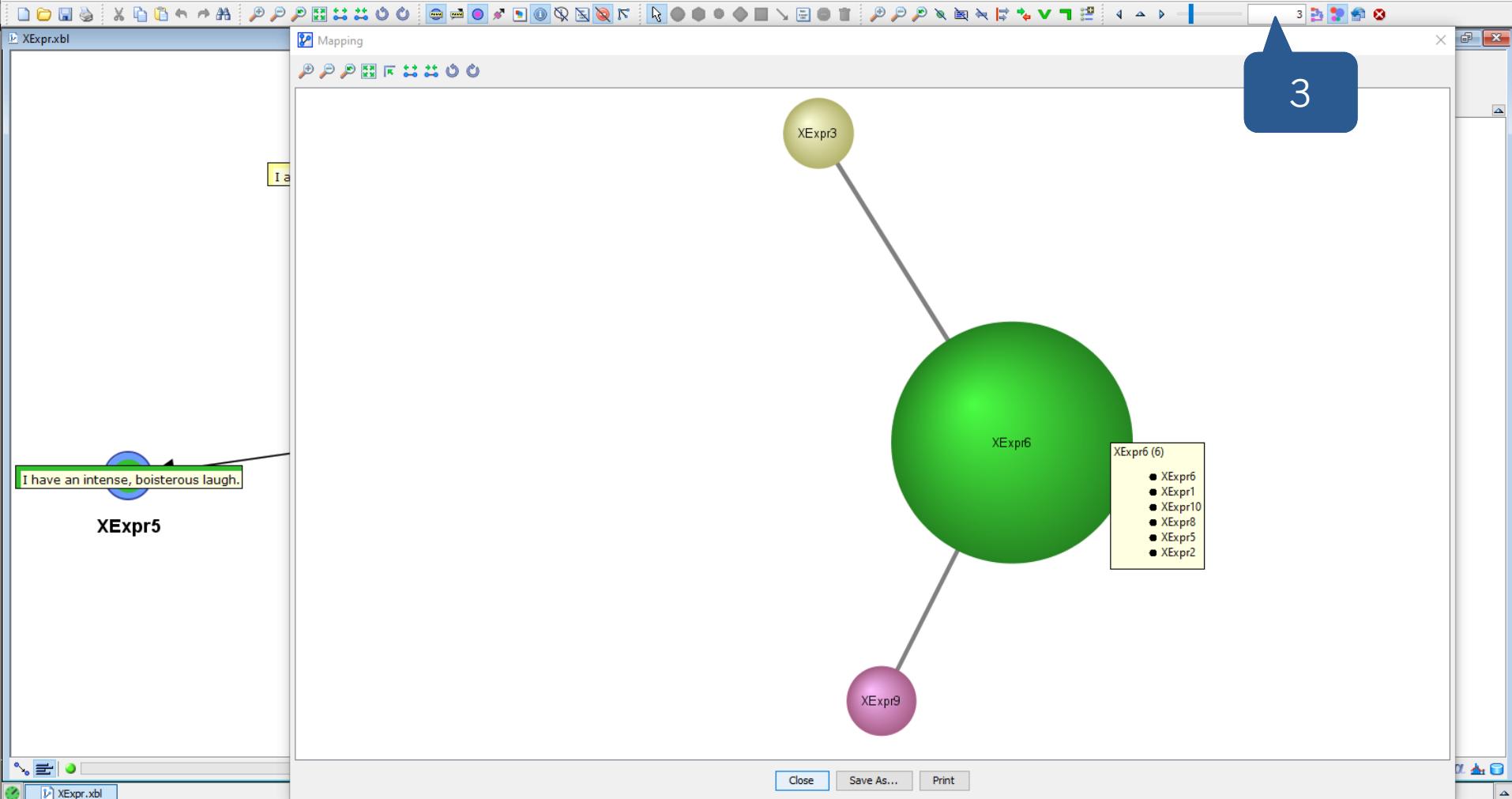


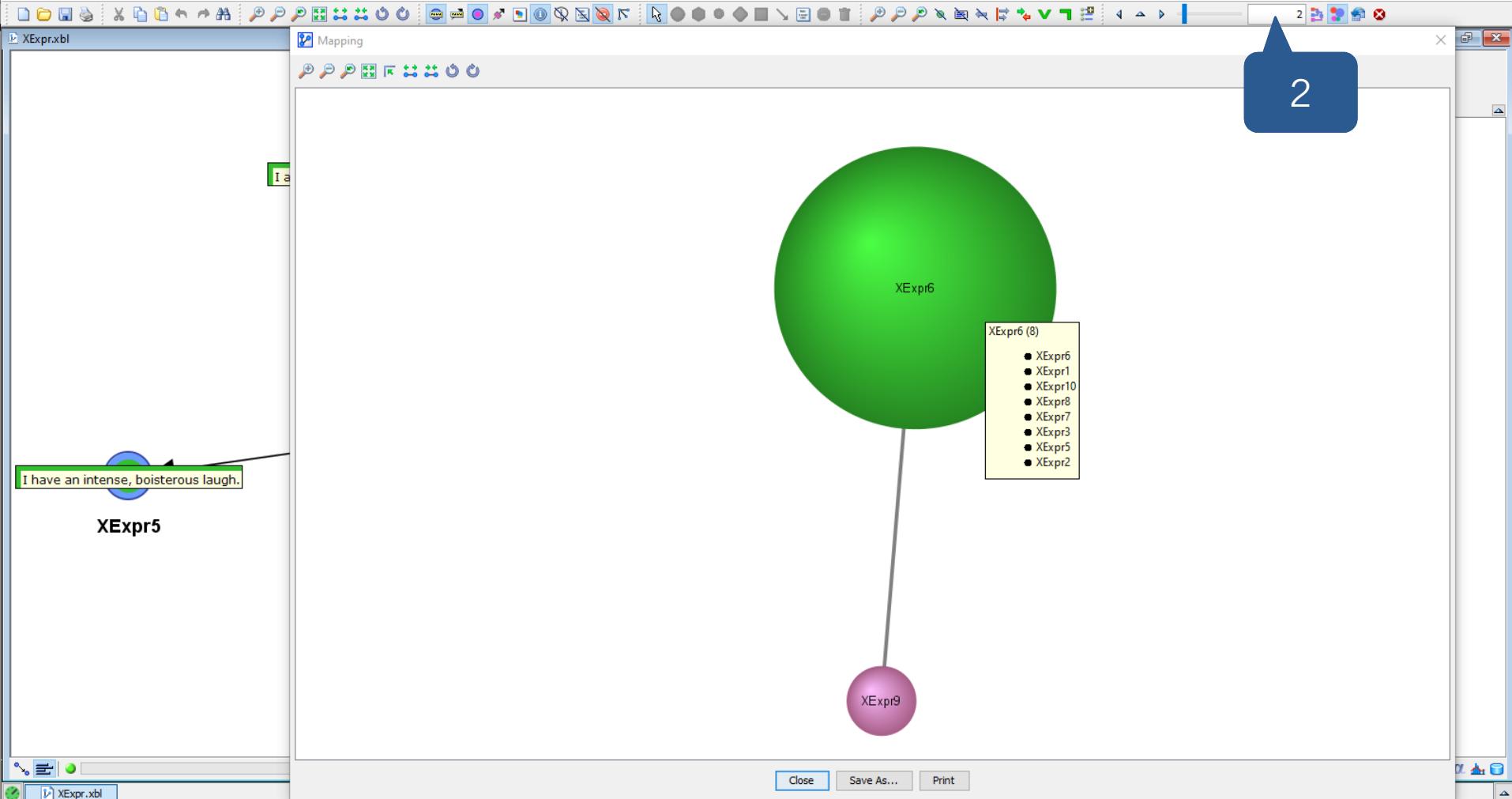


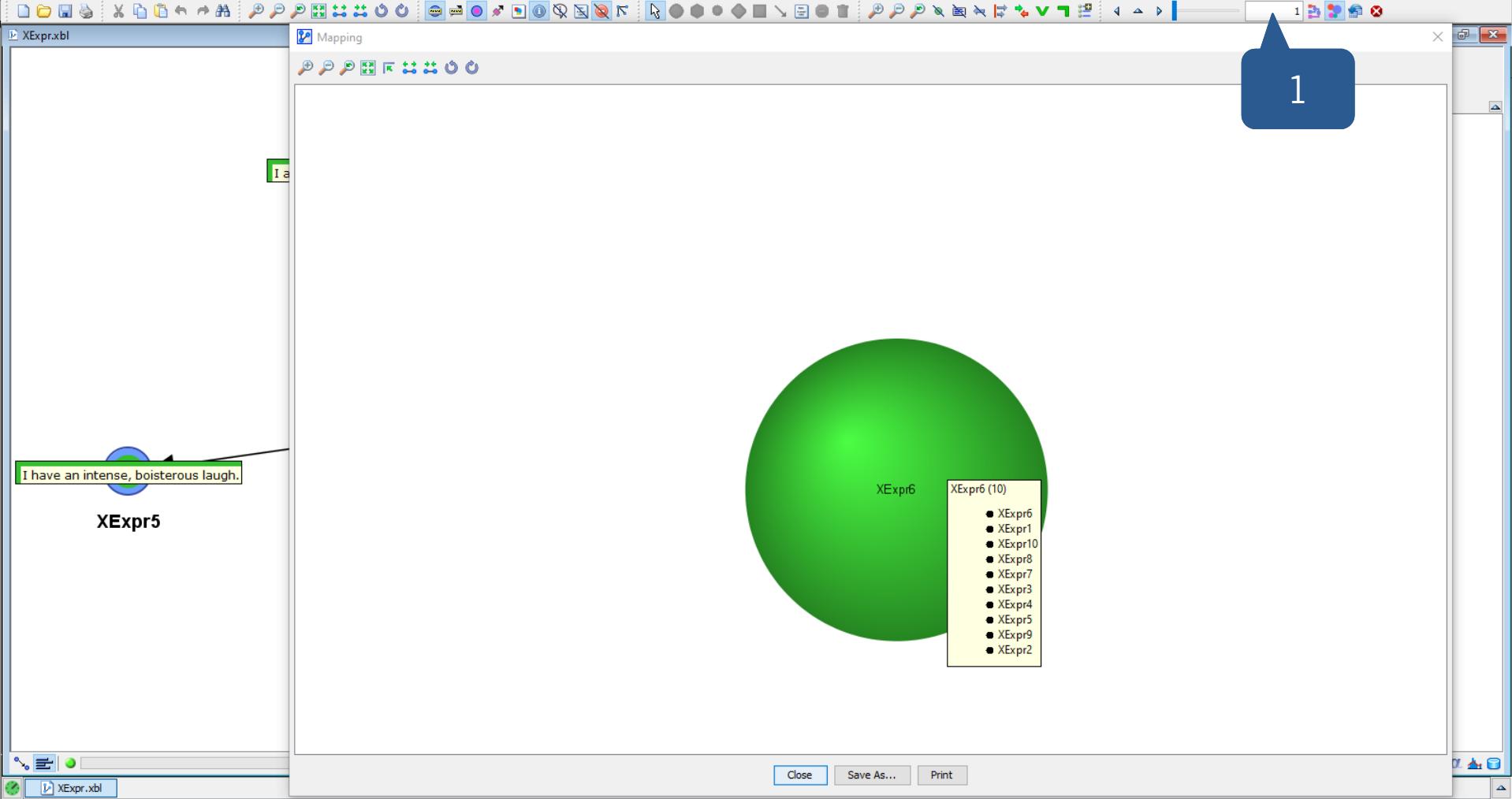












1

I have an intense, boisterous laugh.

XExpr5

I don't talk a lot.

Node Comments

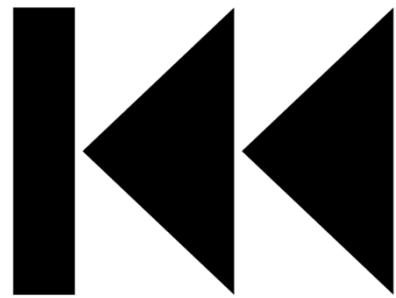
I don't talk a lot. (10)

- I don't talk a lot.
- I talk a lot.
- I speak softly.
- I say little.
- I don't like to draw attention to myself.
- I am the life of the party.
- I tell people about it when i'm irritated.
- I have an intense, boisterous laugh.
- I bottle up my feelings.
- I am never at a loss for words.



But we need
a Bayesian
network
first, right?

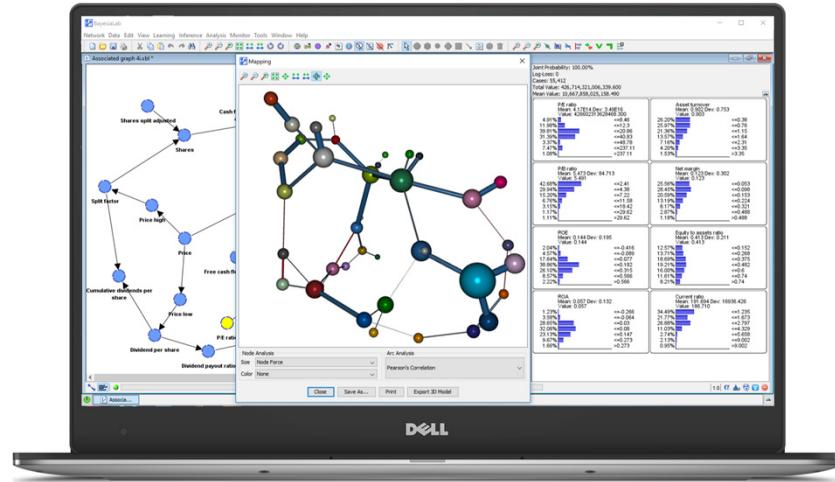
Rewind



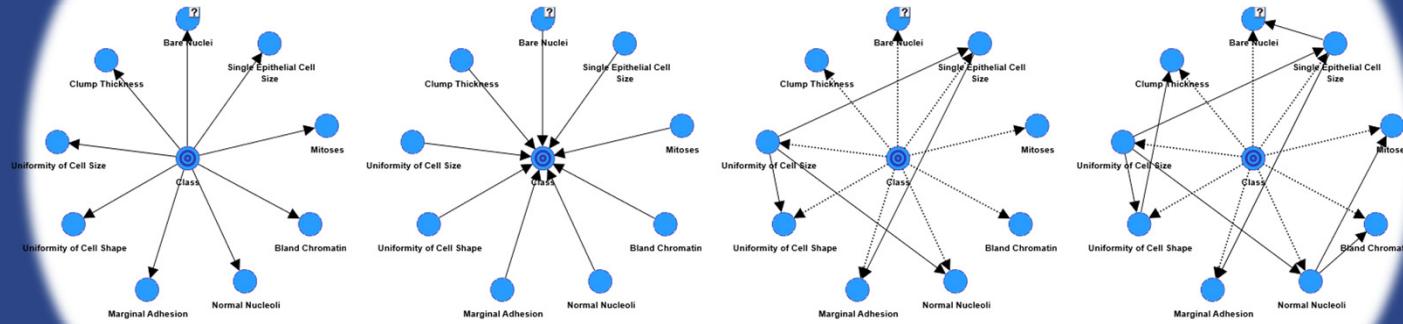
BayesiaLab Workflow

Workflow

- Machine-Learning
 - Minimum Description Length
- Clustering
 - Arc Force
- Validation
 - Log Loss
 - Contingency Table Fit
 - Purity
- Visualization & Interpretation



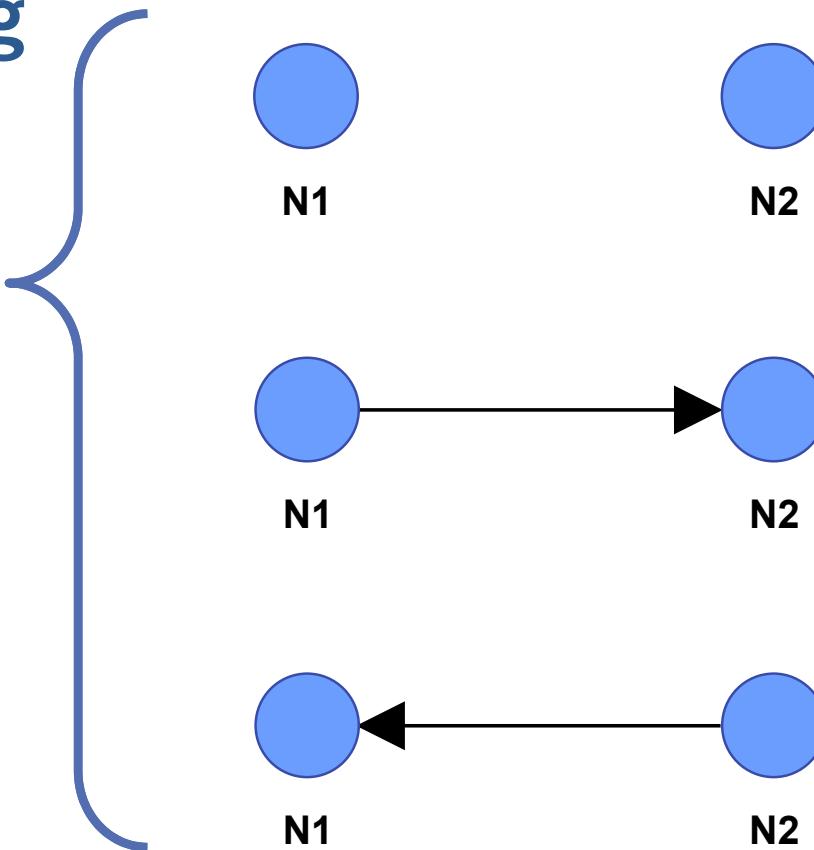
Learning=Searching



Learning=Searching

Number of Possible Networks

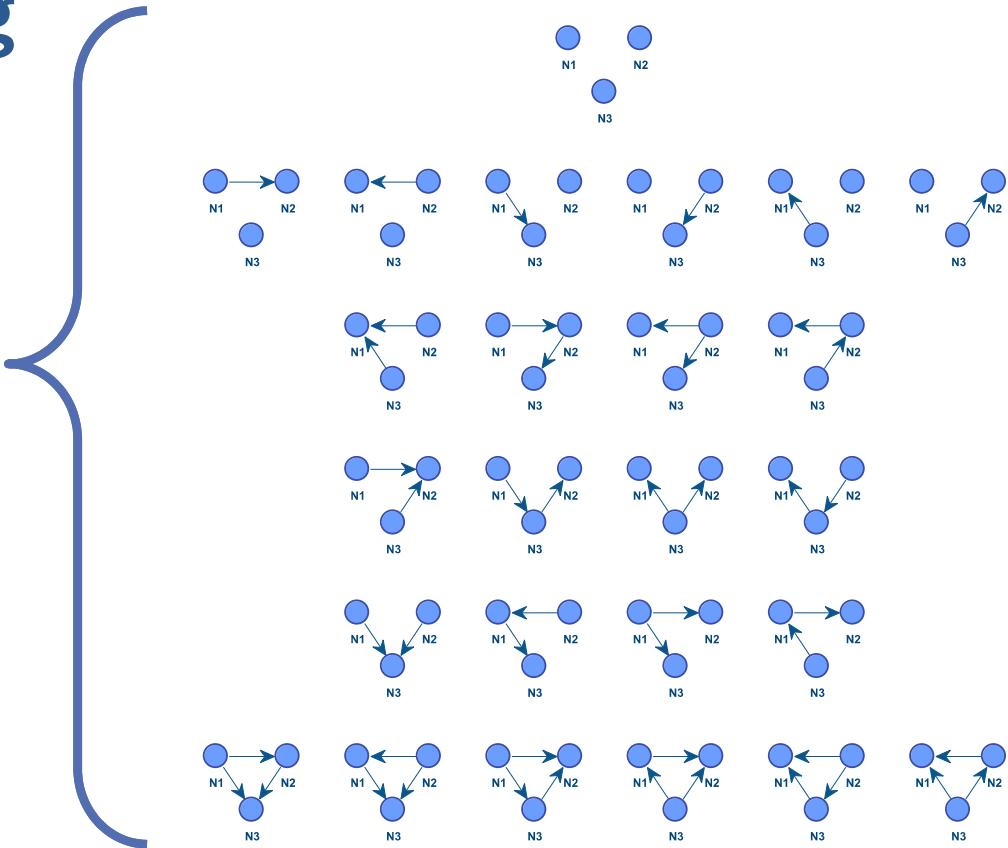
- 2 Nodes: 3



Learning=Searching

Number of Possible Networks

- 2 Nodes: 3
- 3 Nodes: 25



Learning=Searching

Number of Possible Bayesian Networks

- 2 Nodes: 3
- 3 Nodes: 25
- 4 Nodes: 543
- 5 Nodes: 29,281
- 6 Nodes: 3.8×10^6
- 7 Nodes: 1.1×10^9
- 8 Nodes: 7.8×10^{11}
- 9 Nodes: 1.2×10^{15}
- 10 Nodes: 4.2×10^{18}
- 11 Nodes: 3.2×10^{22}
- ⋮ ⋮
- 240 Nodes: 9.1×10^{9060}

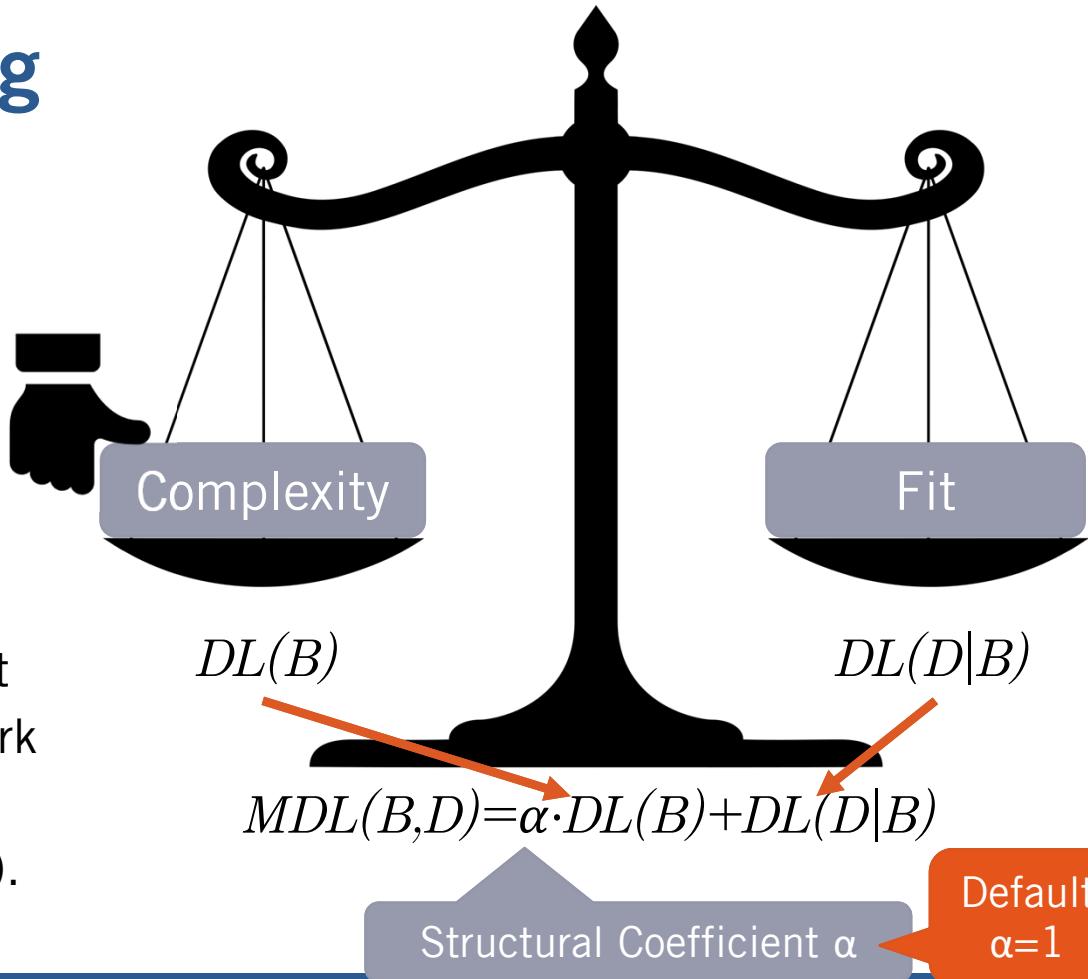
Search Space



Learning=Searching

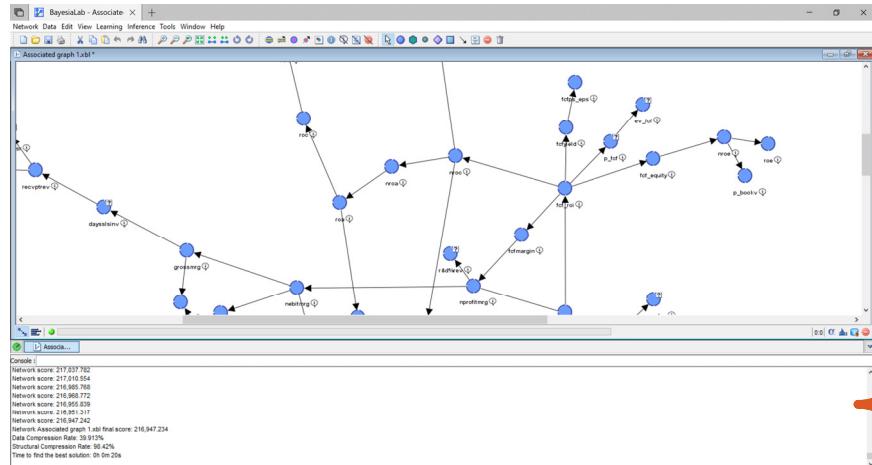
Minimum Description Length

- $DL(B)$ is the number of bits to represent the Bayesian network B (graph and probabilities), and
- $DL(D|B)$ is the number of bits to represent the dataset D given the Bayesian network B (likelihood of the data given the Bayesian network).



Learning=Searching

Minimum Description Length



MDL Score

Network score: 217,884.553
Network score: 217,743.338
Network score: 217,610.856
Network score: 217,483.237
Network score: 217,359.875
Network score: 217,241.952
Network score: 217,195.628
Network score: 217,152.903
Network score: 217,113.827
Network score: 217,075.16
Network score: 217,037.782
Network score: 217,010.554
Network score: 216,985.768
Network score: 216,968.772
Network score: 216,955.839
Network score: 216,951.317
Network score: 216,947.242
Network Associated graph 1.xaml final score: 216,947.234
Data Compression Rate: 39.91%
Structural Compression Rate: 98.42%
Time to find the best solution: 0h 0m 20s



Data Import

Define Data Structure

Separators

Tab Semicolon Comma
 Space Other

Encoding

UTF-8

Missing Values

N/R
NR
NC

Filtered Values

VF
FV
N/A

Sampling

Define Sample

Learning/Test

Define Learning/Test Sets

Options

Title Line
 End of Line Character
 Consider Identical Consecutive separators as a Unique One
 Consider Different Consecutive Separators as a Unique One
 Double Quote as String Delimiters
 Single Quote as String Delimiters
 Transpose

Data

HSinc1	HSinc2	HSinc3	HSinc4	HSinc5	HSinc6	HSinc7	HSinc8	HSinc9	HSinc10
2	3	5	3	2	3	3	6	3	2
4	4	5	5	4	6	6	6	4	3
6	2	3	3	2	5	2	5	5	6
1	5	6	5	7	7	6	7	7	7
6	4	5	4	5	5	4	5	5	4
6	3	3	6	3	2	1	7	3	2
3	1	6	6	4	4	4	6	5	7
1	3	5	1	3	2	1	5	5	2
5	3	5	7	4	1	6	5	1	5
5	3	4	4	4	4	1	6	3	3
3	1	3	2	6	2	5	6	4	1
6	4	4	2	1	3	2	5	5	1

Cancel Previous Next Save Finish



Data Import

Define Variable Type

Type	Action	Information
<input type="radio"/> Not Distributed	Columns with Missing Values	Number of Rows 15017 100.00%
<input checked="" type="radio"/> Discrete	All not Distributed	Not Distributed 0 0.00%
<input type="radio"/> Continuous	All Discrete	Discrete 240 100.00%
<input type="radio"/> Weight	All Continuous	Continuous 0 0.00%
<input type="radio"/> Learning/Test		Others 0 0.00%
<input type="radio"/> Row Identifier		Missing Values 0 0.00%
		Filtered Values 0 0.00%

Data

HSinc1	HSinc2	HSinc3	HSinc4	HSinc5	HSinc6	HSinc7	HSinc8	HSinc9	HSinc10
2	3	5	3	2	3	3	6	3	2
4	4	5	5	4	6	6	6	4	3
6	2	3	3	2	5	2	5	5	6
1	5	6	5	7	7	6	7	7	7
6	4	5	4	5	5	4	5	5	4
6	3	3	6	3	2	1	7	3	2
3	1	6	6	4	4	4	6	5	7
1	3	5	1	3	2	1	5	5	2
5	3	5	7	4	1	6	5	1	5
5	3	4	4	4	4	1	6	3	3
3	1	3	2	6	2	5	6	4	1
6	4	4	2	1	3	2	5	5	1
7	2	1	2	1	1	1	3	4	1
4	2	3	2	7	2	1	4	1	1

Cancel Previous Next Save Finish



Data Import

Data Selection and Filtering

Missing Value Processing

Filter
 OR
 AND
 Replace by :
 Value
 Mean/Modal
 Infer
 Static Imputation
 Dynamic Imputation
 Structural EM
 Entropy-Based Static Imputation
 Entropy-Based Dynamic Imputation

Information

Number of Rows	15017	100.00%
Not Distributed	0	0.00%
Discrete	240	100.00%
Continuous	0	0.00%
Others	0	0.00%
Missing Values	0	0.00%
Filtered Values	0	0.00%

Select Values

OR
 AND

Data

HSinc1	HSinc2	HSinc3	HSinc4	HSinc5	HSinc6	HSinc7	HSinc8	HSinc9	HSinc10
2	3	5	3	2	3	3	6	3	2
4	4	5	5	4	6	6	6	4	3
6	2	3	3	2	5	2	5	5	6
1	5	6	5	7	7	6	7	7	7
6	4	5	4	5	5	4	5	5	4

Cancel Previous Next Save Finish



Data Import

Discretization and Aggregation

Aggregation

States (7)
1
2
3
4
5
6
7

Aggregates (0)

Display Correlations

Target: AFlex1

State:

Automatic Aggregation

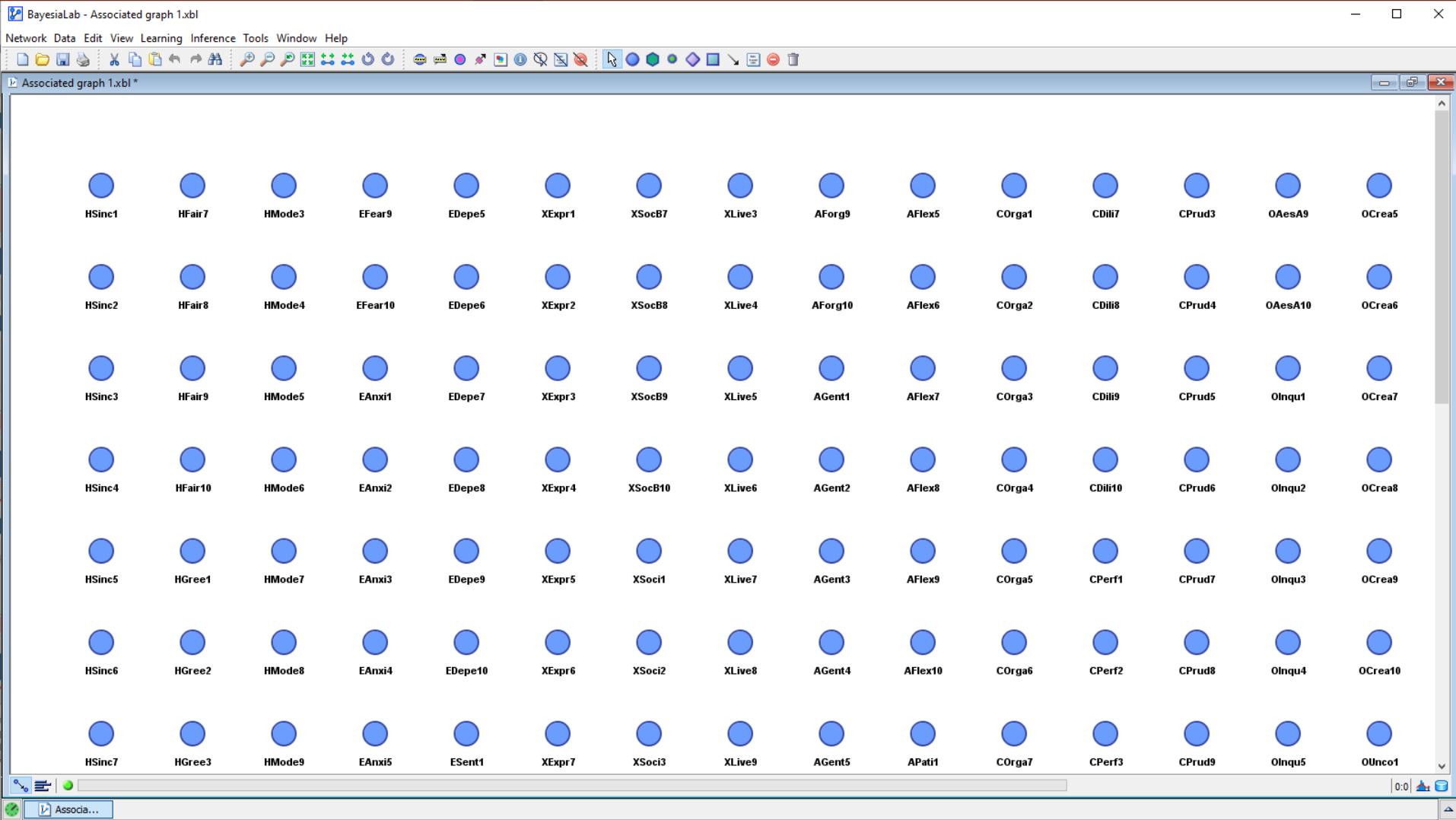
Transfer Aggregates

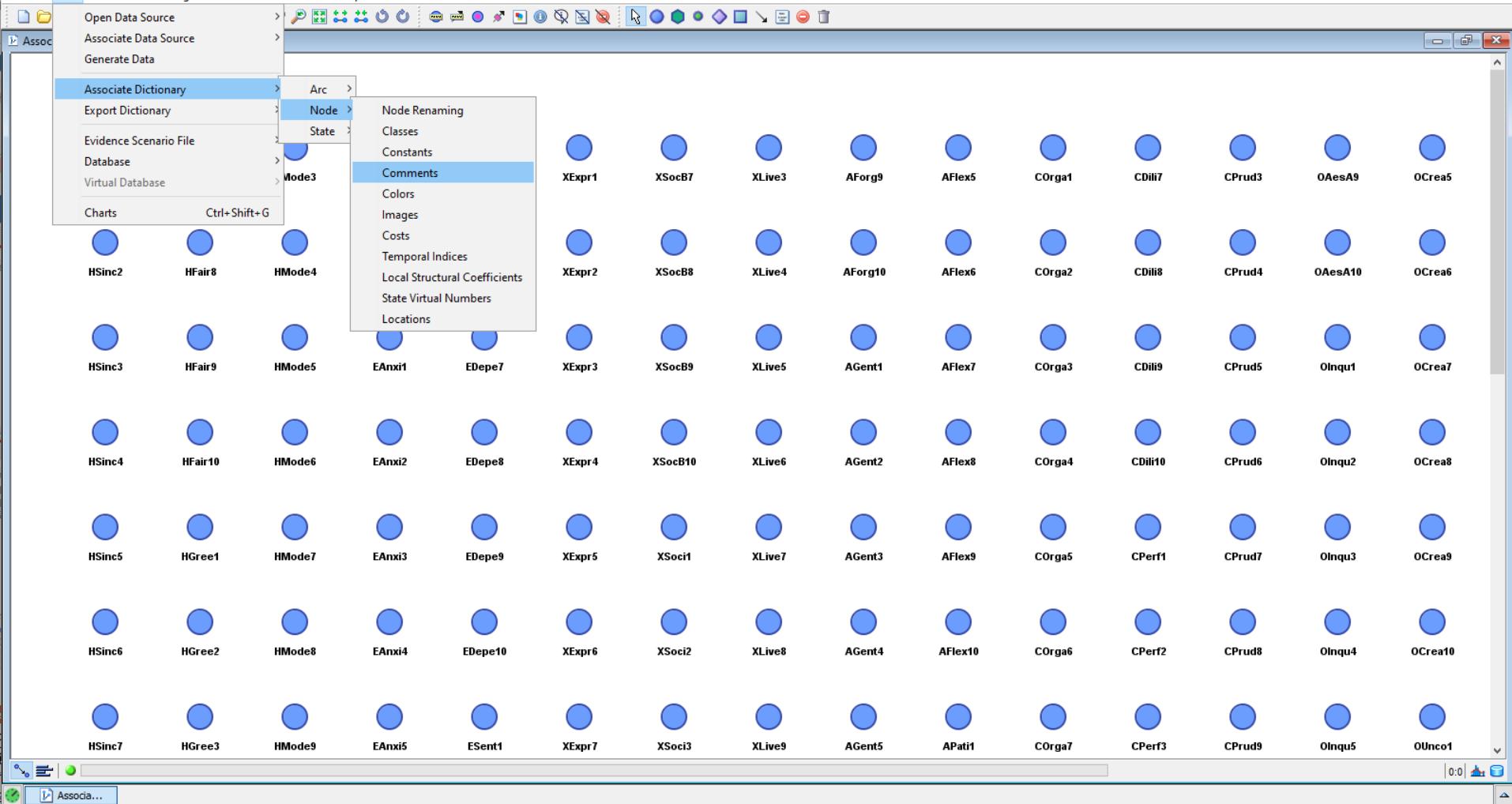
Data

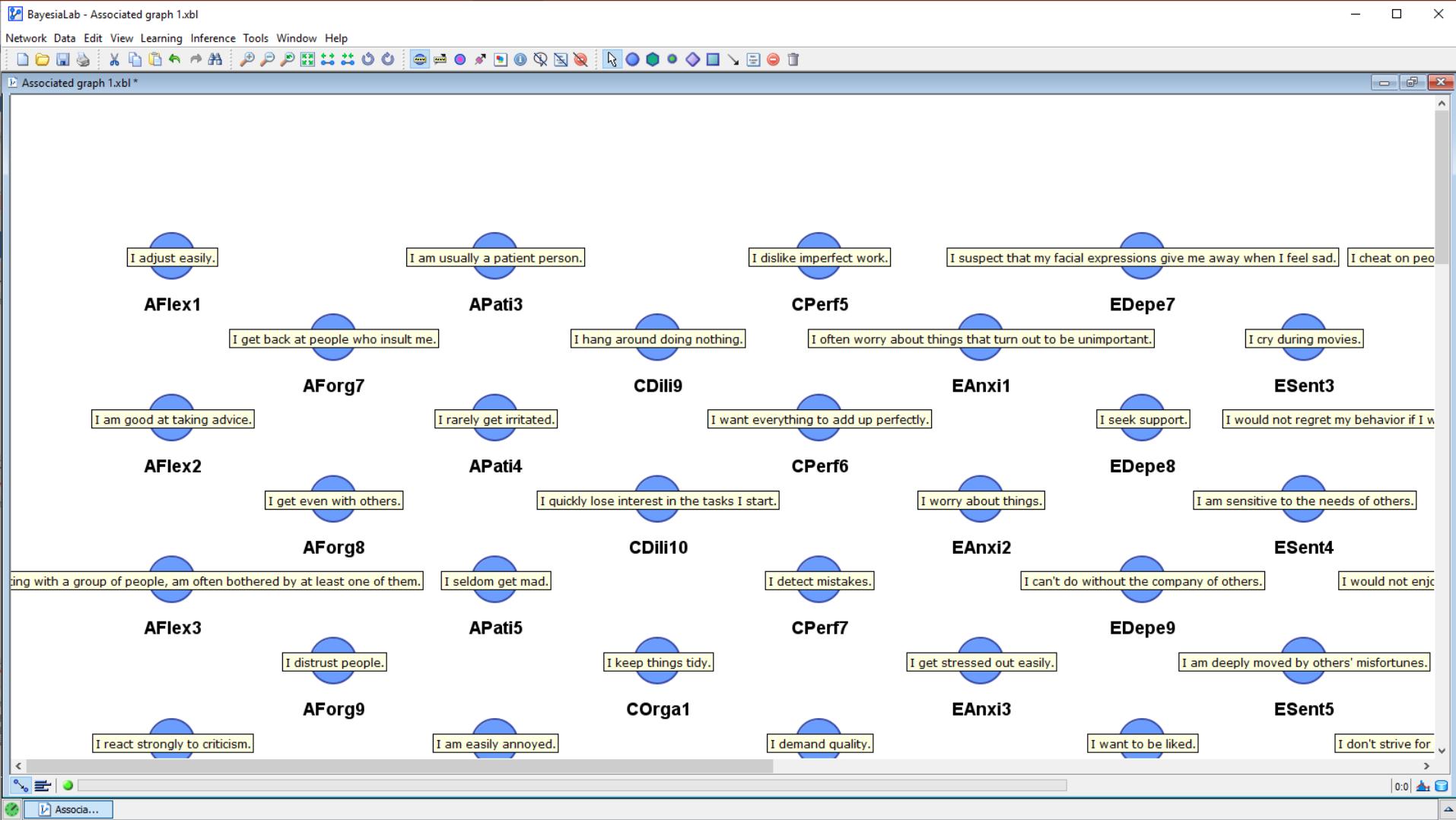
HSinc1	HSinc2	HSinc3	HSinc4	HSinc5	HSinc6	HSinc7	HSinc8	HSinc9	HSinc10
2	3	5	3	2	3	3	6	3	2
4	4	5	5	4	6	6	6	4	3
6	2	3	3	2	5	2	5	5	6
1	5	6	5	7	7	6	7	7	7
6	4	5	4	5	5	4	5	5	4

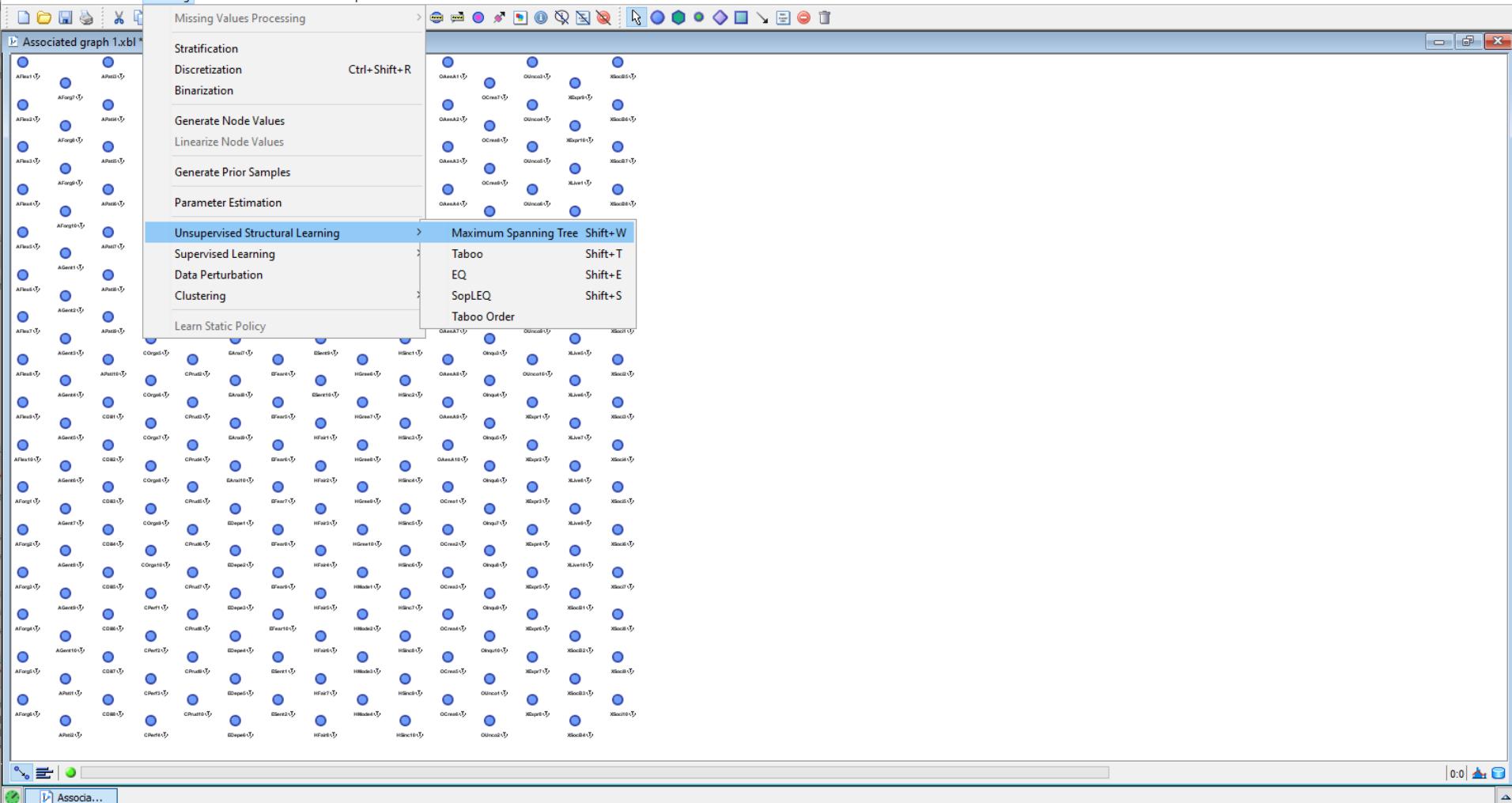
Select All Continuous Select All Discrete

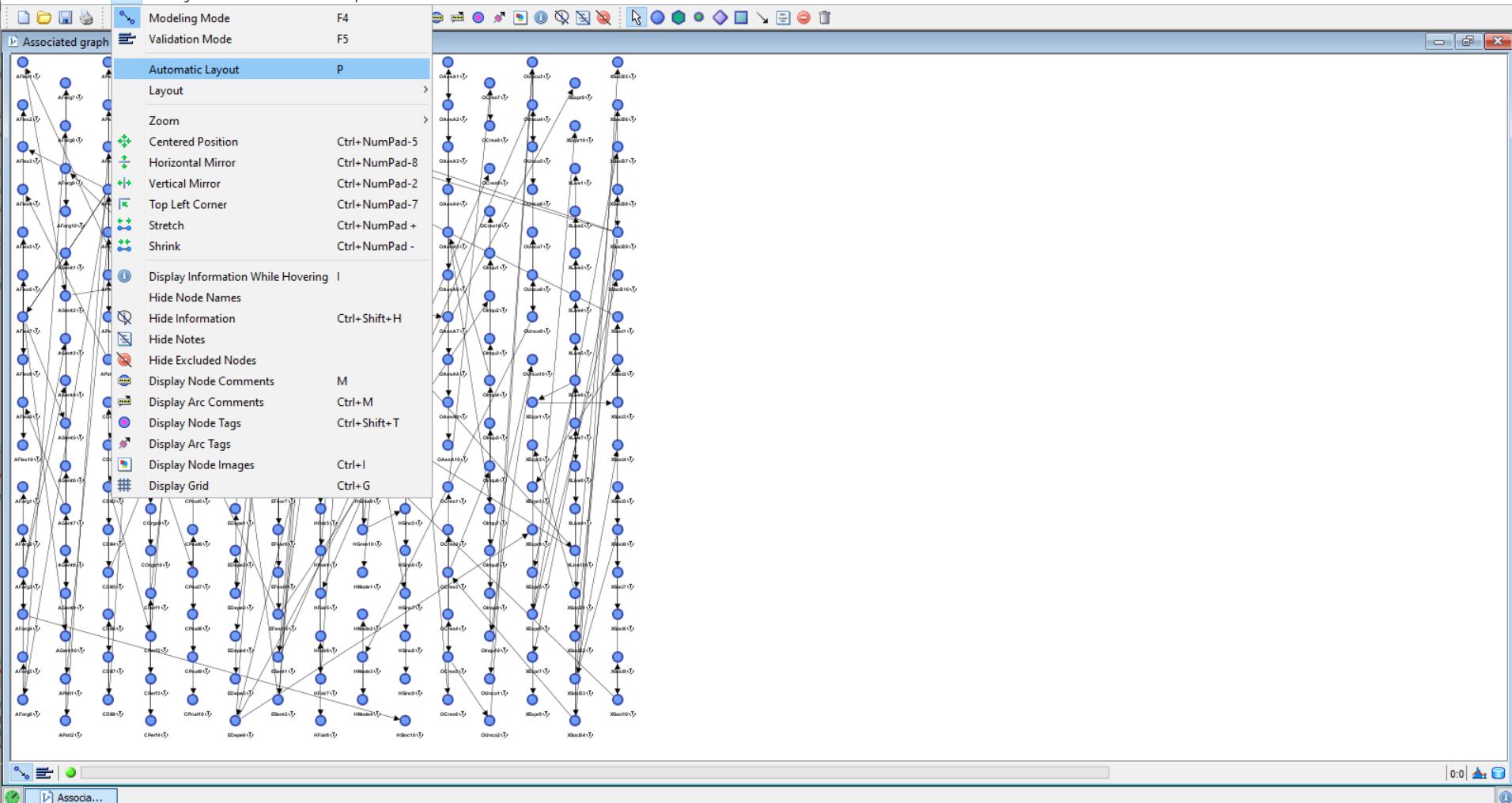
Cancel Previous Next Save Finish





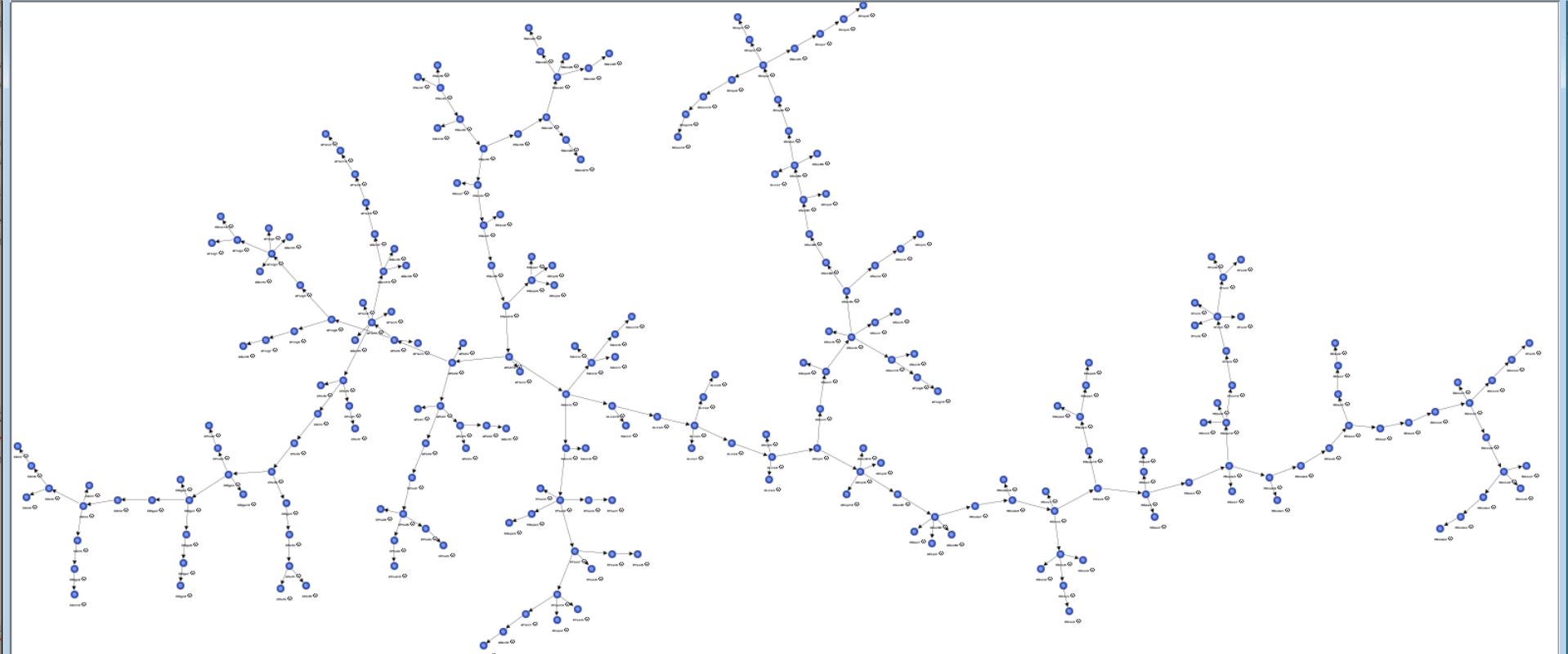


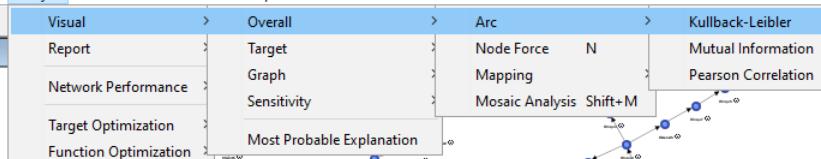
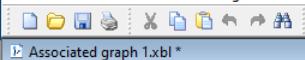




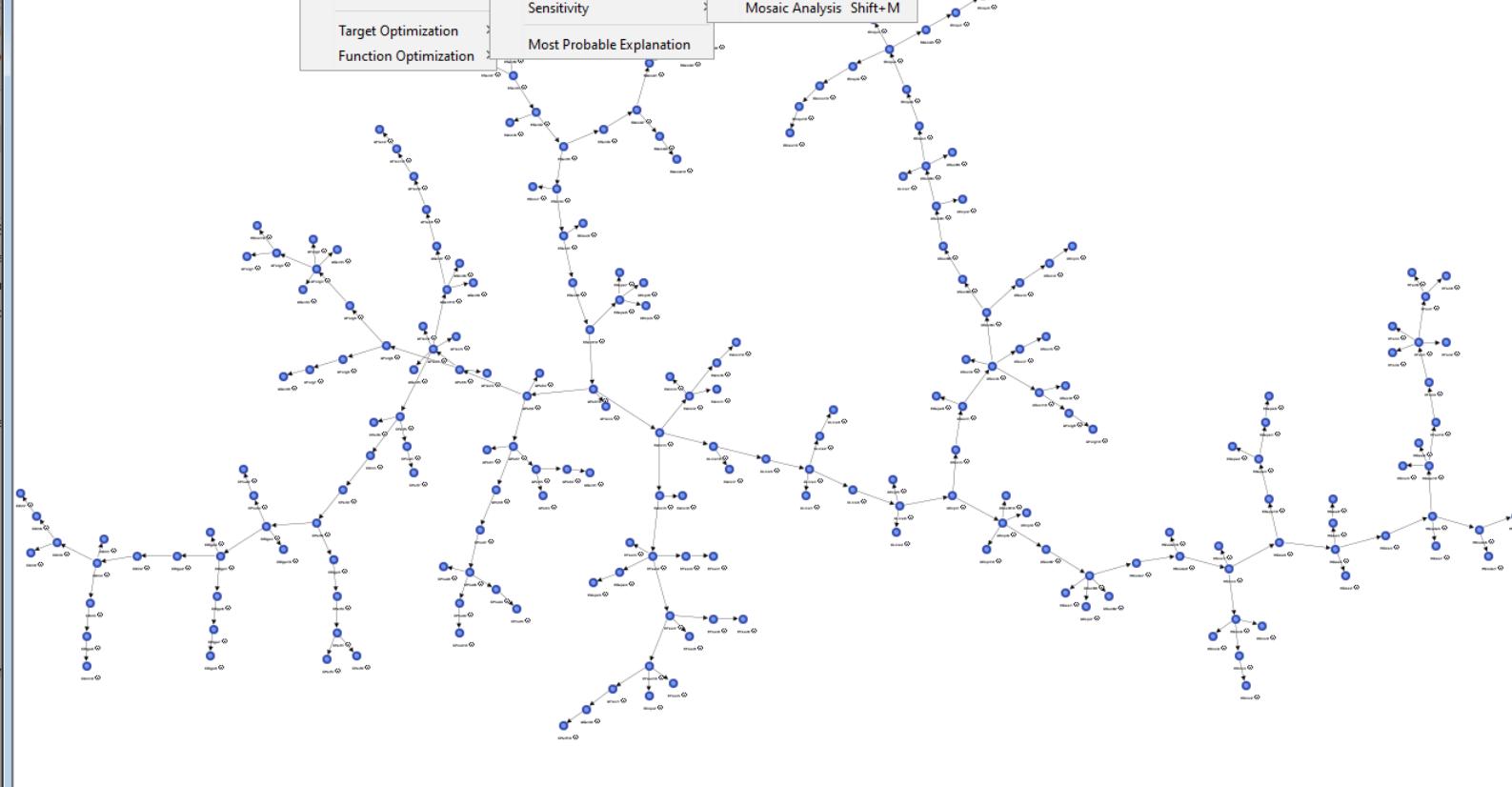


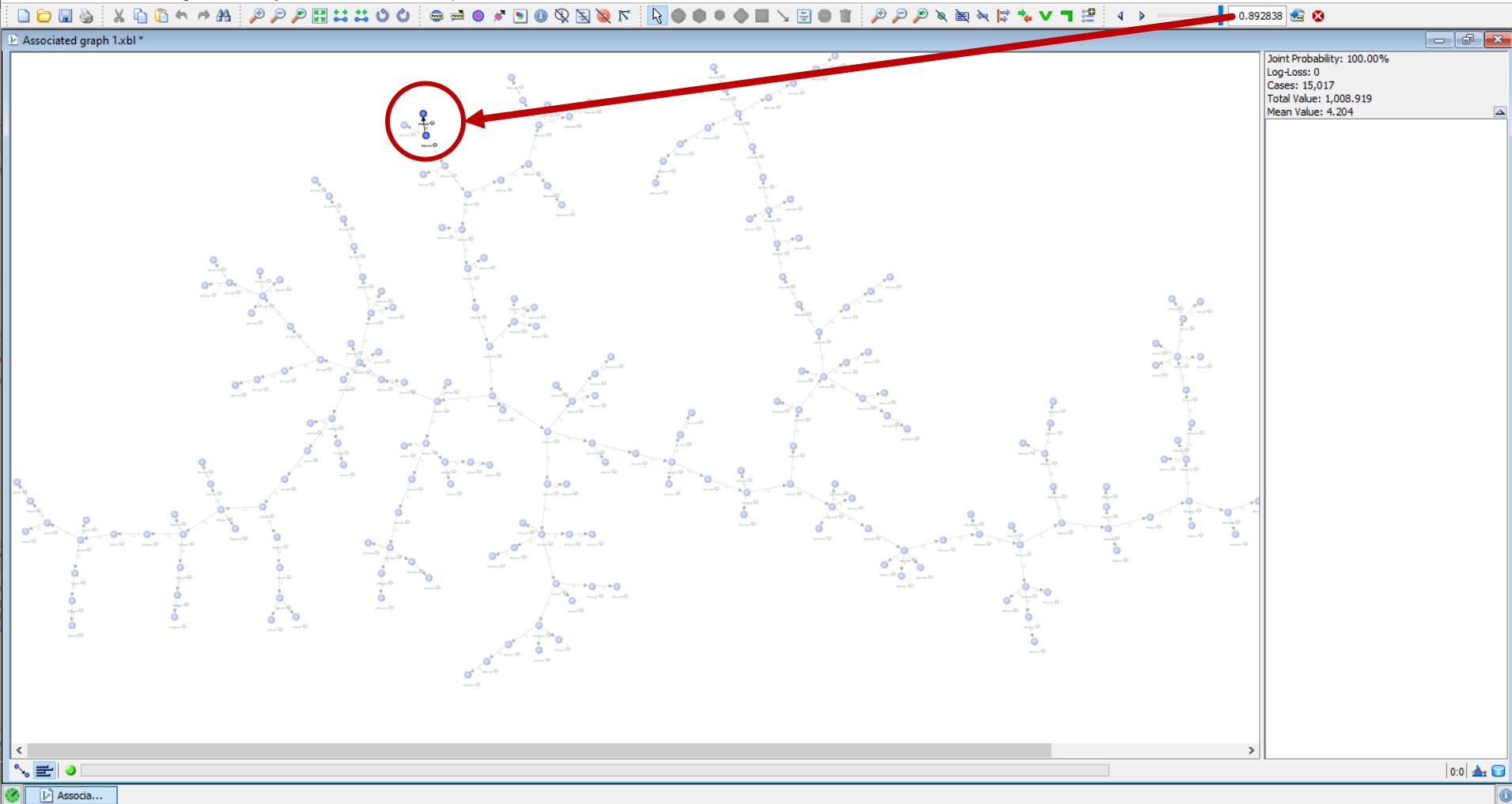
Associated graph 1.xaml *

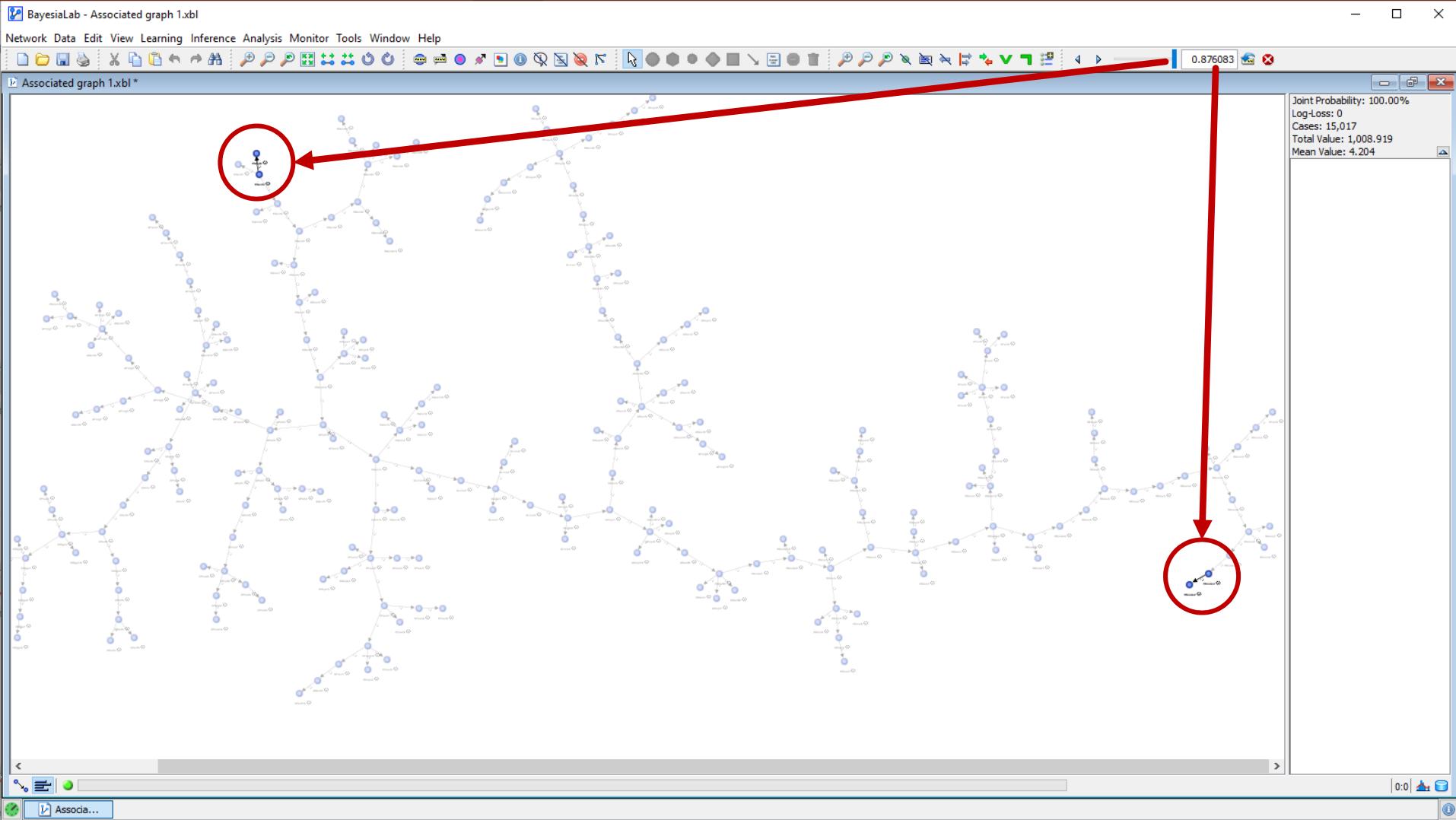


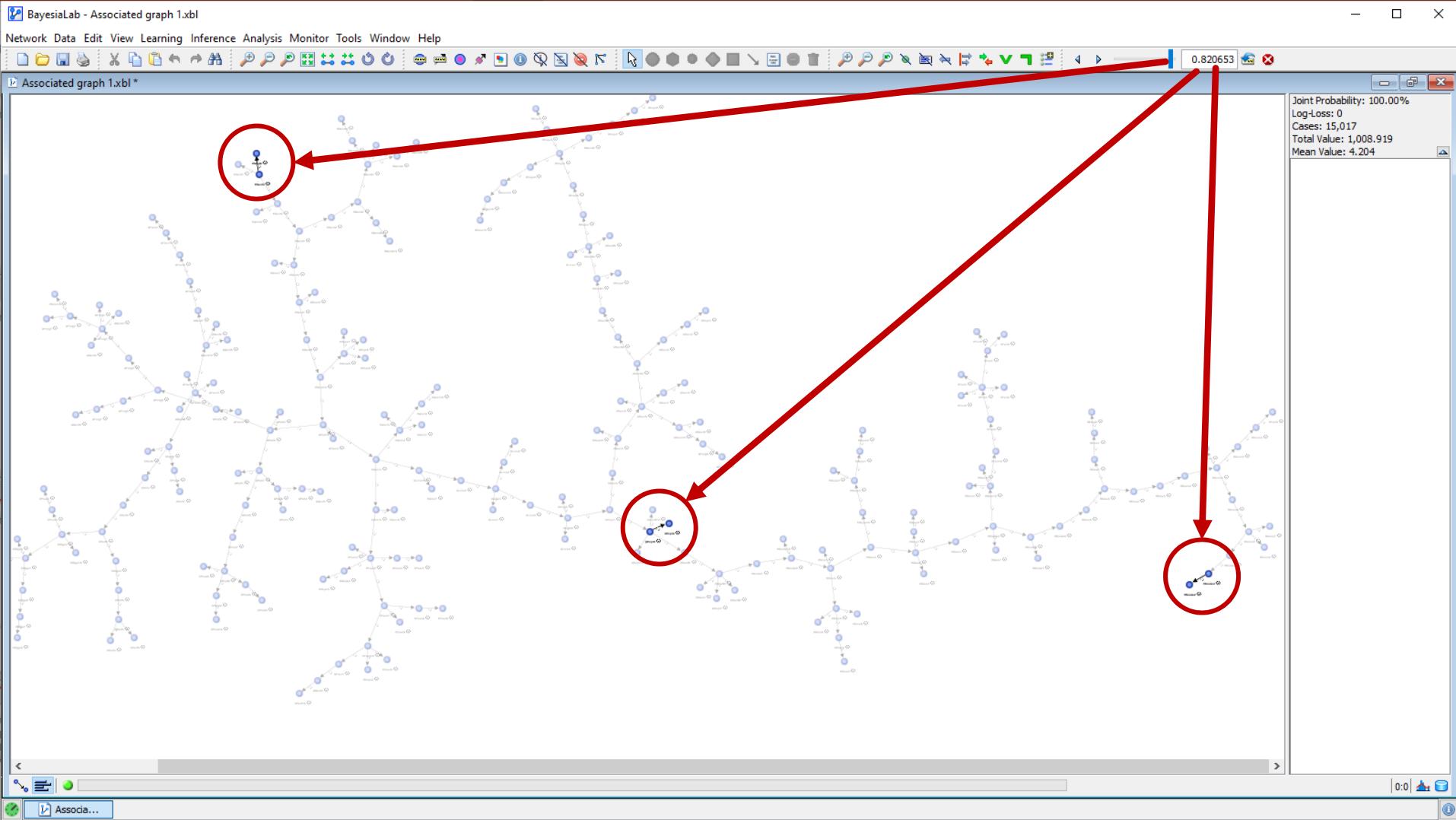


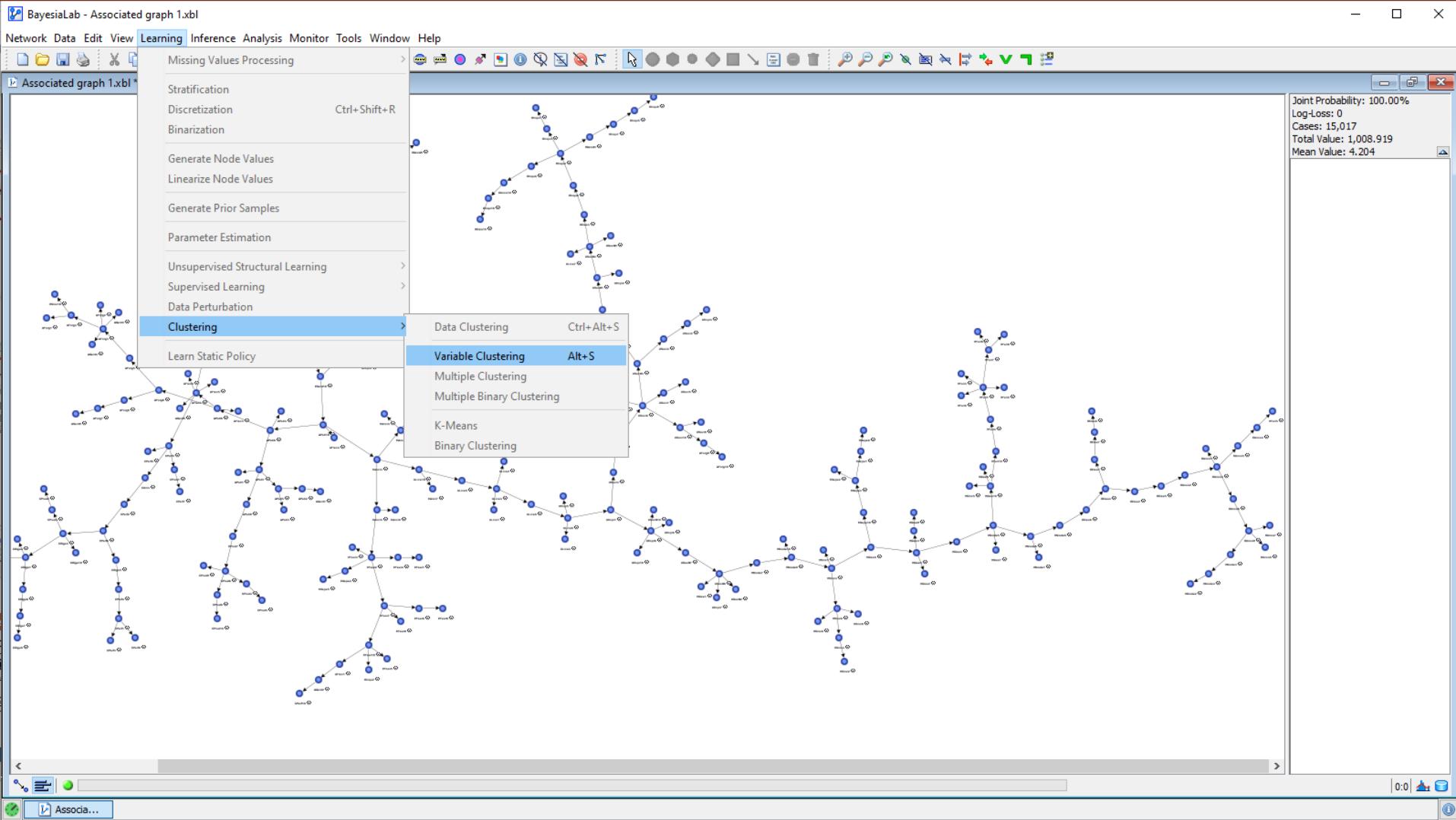
Joint Probability: 100.00%
Log-Loss: 0
Cases: 15,017
Total Value: 1,008.919
Mean Value: 4.204

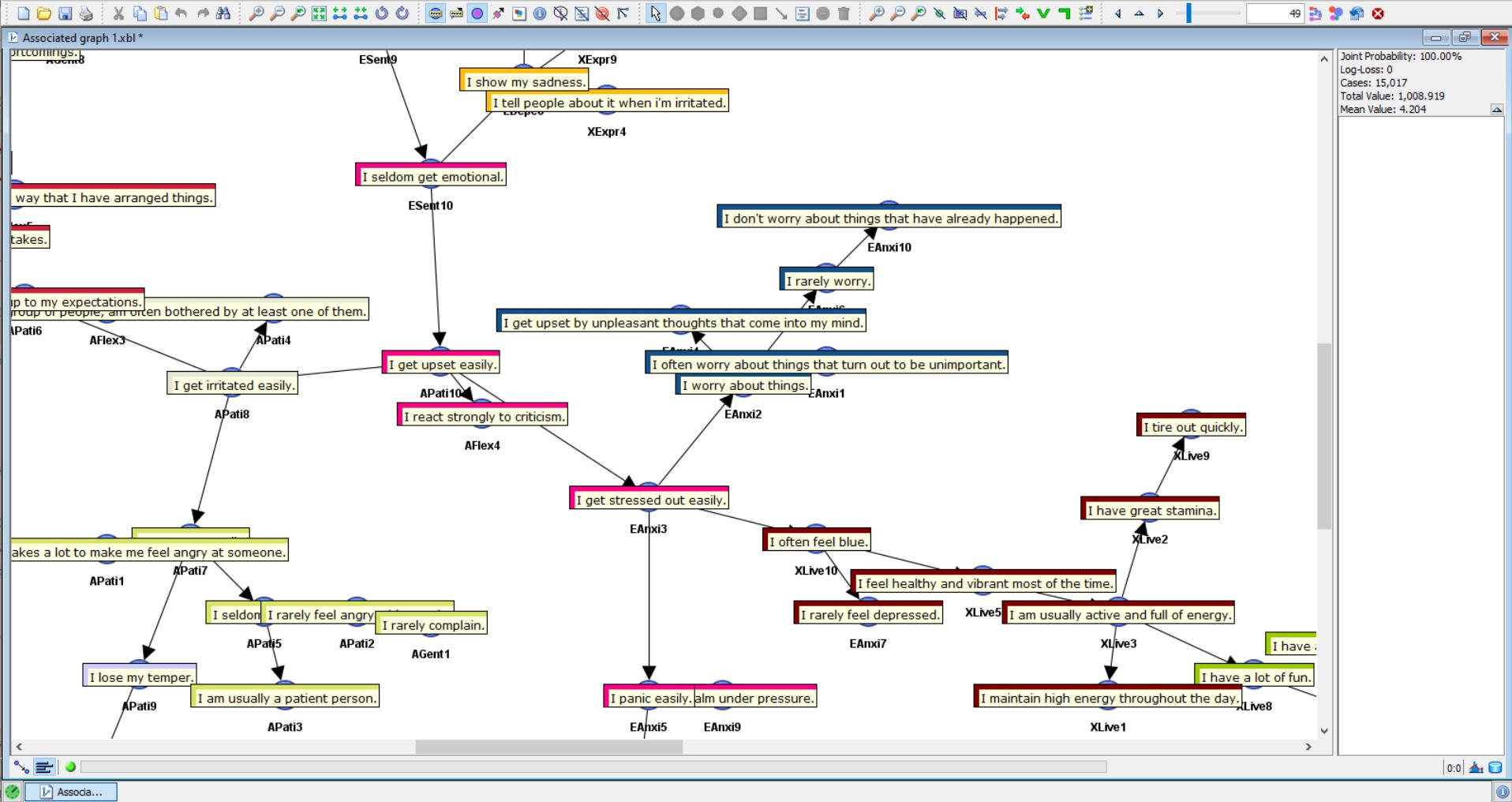


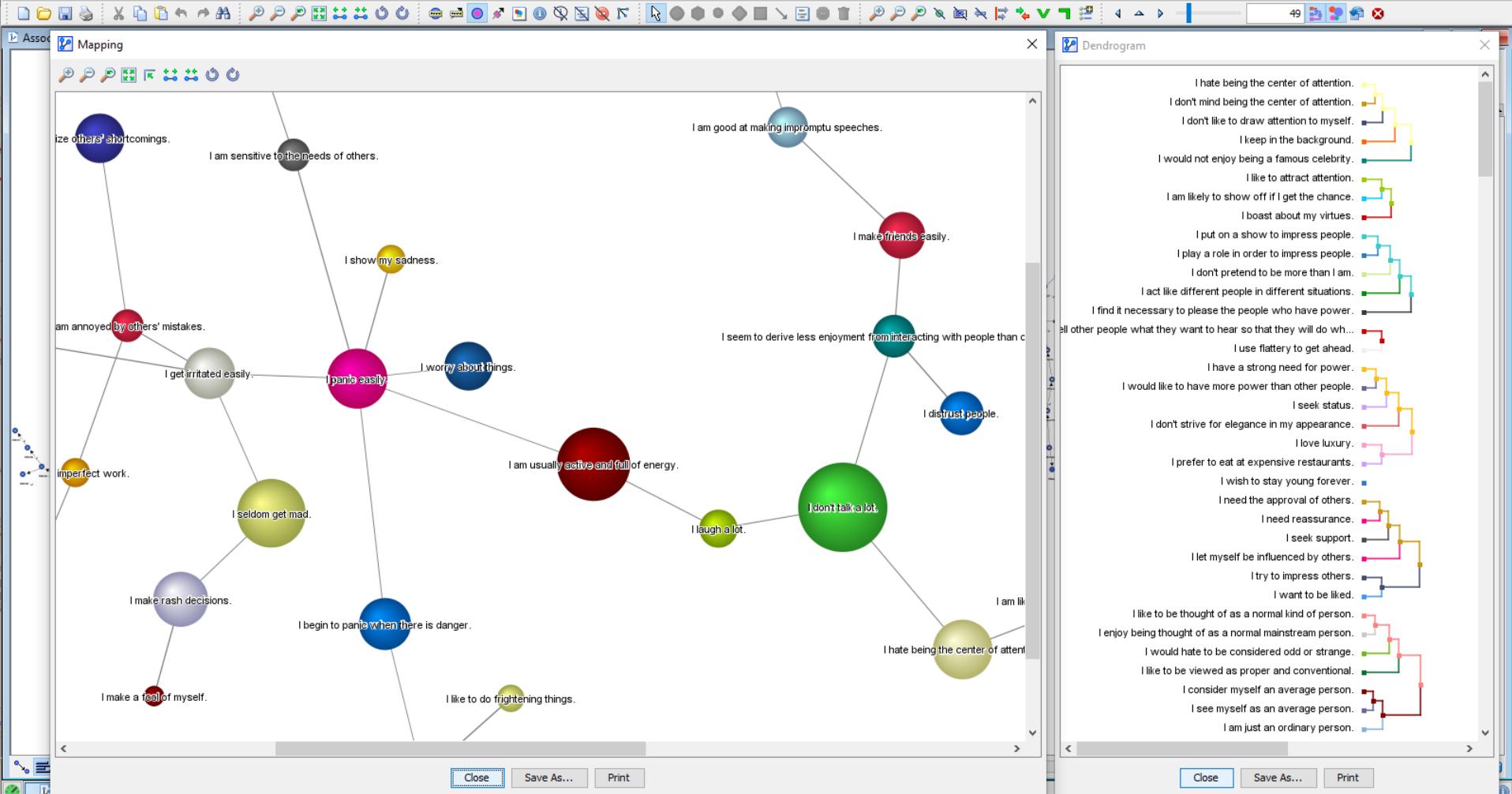




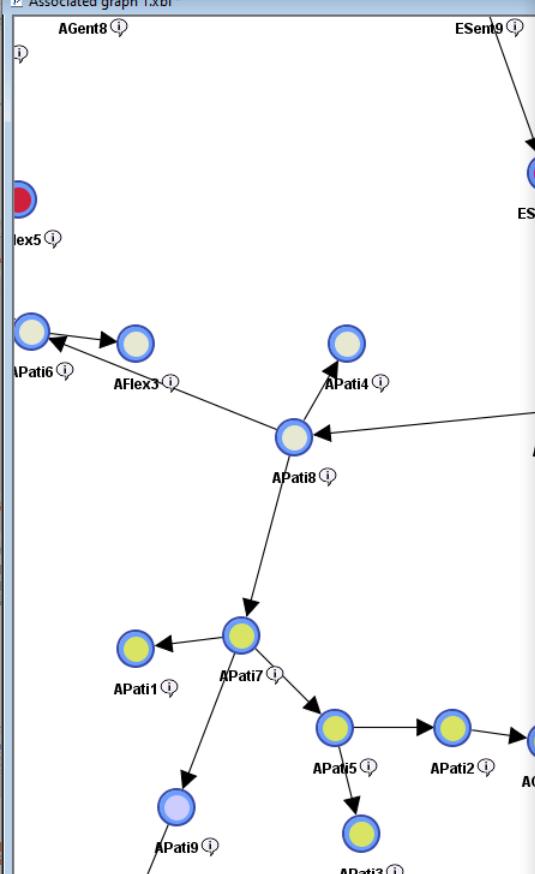








Associated graph 1.xaml *



Classes	Nodes	Comments
[Factor_0]	ESent3	I cry during movies.
	ESent6	I rarely cry during sad movies.
	ESent2	I immediately feel sad when hearing of an unhappy event.
	ESent5	I am deeply moved by others' misfortunes.
	ESent7	I seldom feel weepy while reading the sad part of a story.
	ESent8	I am seldom bothered by the apparent suffering of strangers.
[Factor_1]	EAnxi8	I am not easily disturbed by events.
	XSocB1	I am good at making impromptu speeches.
	XSocB4	I have leadership abilities.
	XSocB6	I would be afraid to give a speech in public.
	XSocB5	I have a strong personality.
	XExpr2	I am never at a loss for words.
[Factor_2]	XLive7	I feel that I have a lot of inner strength.
	EAnxi5	I panic easily.
	APati10	I get upset easily.
	EAnxi3	I get stressed out easily.
	EAnxi9	I remain calm under pressure.
	ESent10	I seldom get emotional.
[Factor_3]	ESent9	I don't understand people who get emotional.
	AFlex4	I react strongly to criticism.
	HMode4	I consider myself an average person.
	OUNco9	I like to be thought of as a normal kind of person.
	HMode3	I am just an ordinary person.
	HMode2	I see myself as an average person.
[Factor_4]	OUNco7	I enjoy being thought of as a normal mainstream person.
	OUNco6	I would hate to be considered odd or strange.
	OUNco8	I like to be viewed as proper and conventional.
	EFear2	I begin to panic when there is danger.
	EFear4	I tremble in dangerous situations.
	EDeve4	I need protection.
[Factor_5]	EFear1	I am a physical coward.
	EDeve5	I often need help.
	EFear3	I would fear walking in a high-crime part of a city.
	APati5	I seldom get mad.
	APati7	I get angry easily.
	APati2	I rarely feel angry with people.
	APati1	I find that it takes a lot to make me feel angry at someone.

Close

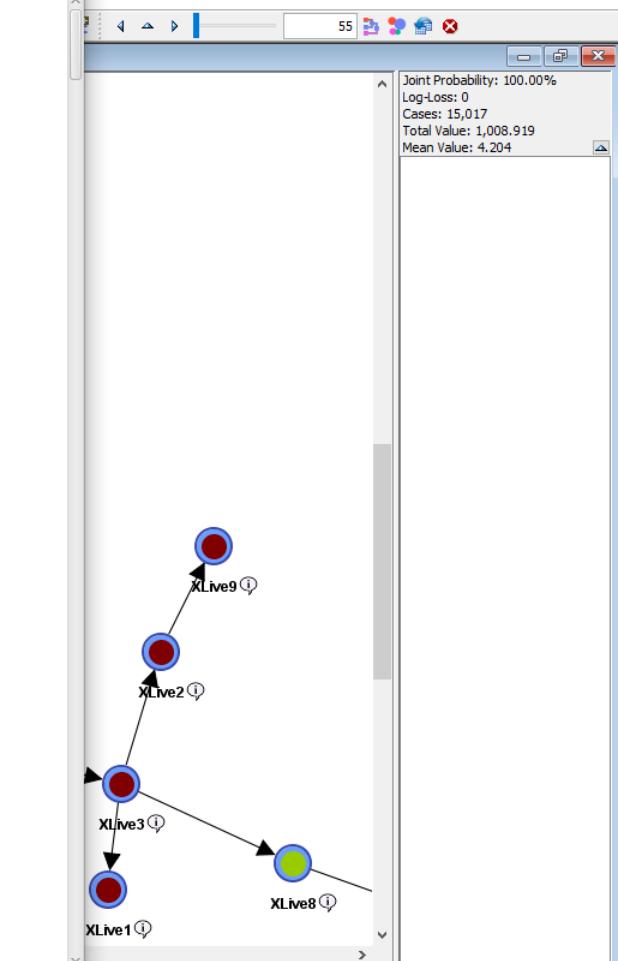
Save As...

Print

55

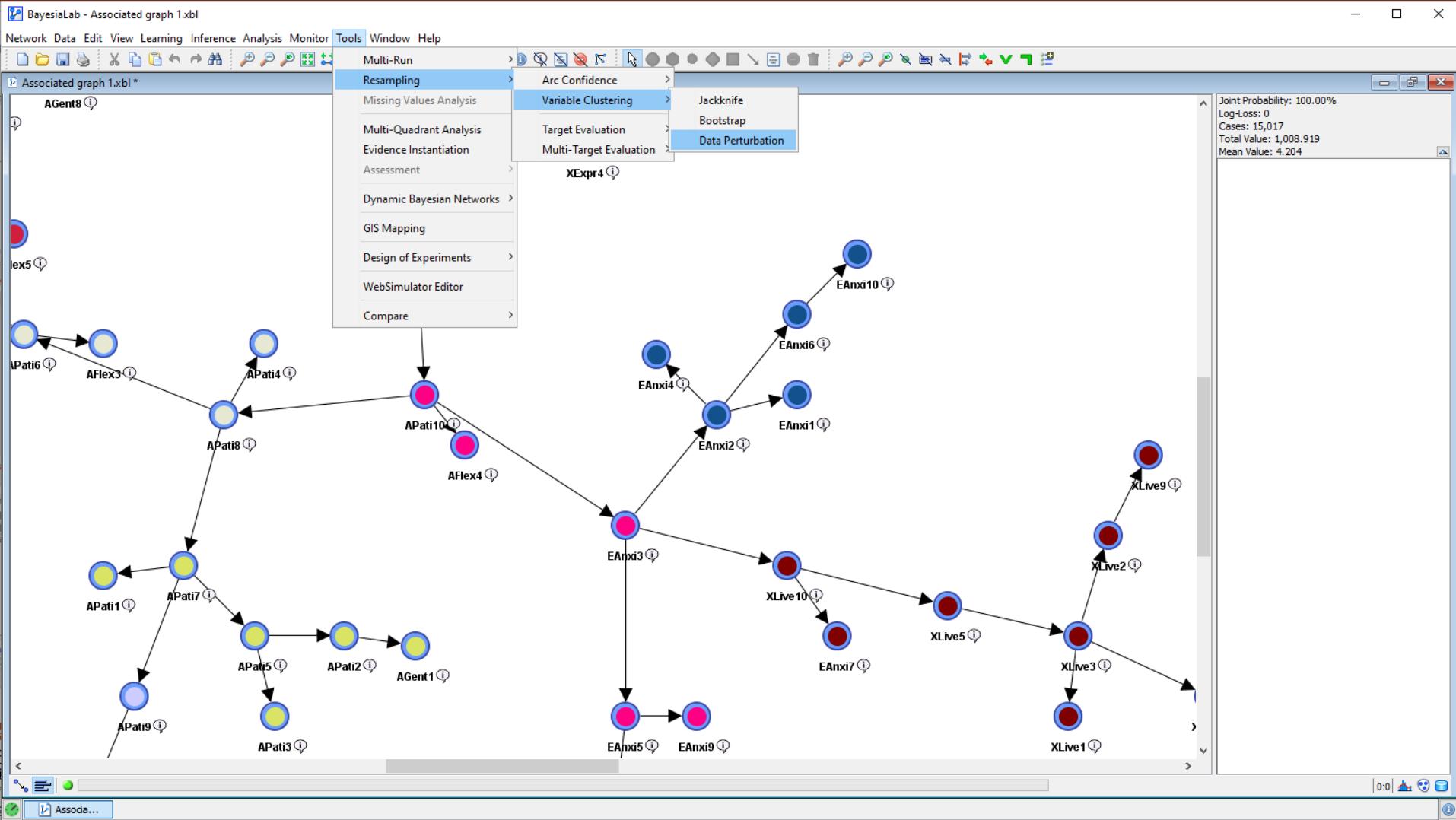


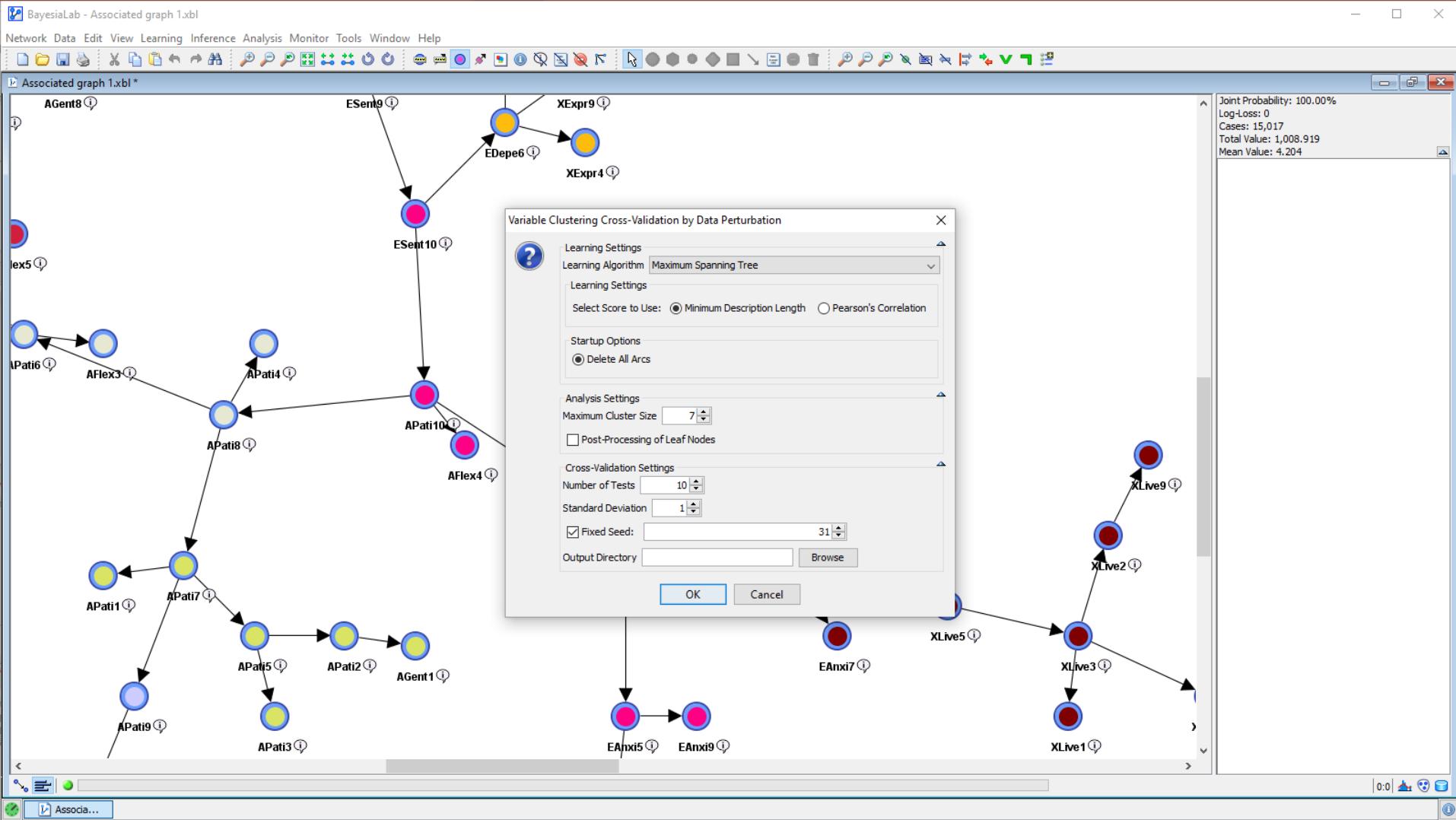
Joint Probability: 100.00%
Log-Loss: 0
Cases: 15,017
Total Value: 1,008.919
Mean Value: 4.204



0:0







BayesiaLab - Associated graph

Variable Clustering Report

XSocB10 I have little to say.

AForg1 I love my enemies.

AForg4 I am nice to people I should be angry at.

HSinc10 I let people push me around to help them feel important.

HFair6 I admire a really clever scam.

HFair7 I cheat to get ahead.

HFair9 I cheat on people who have trusted me.

AFlex1 I adjust easily.

AGent2 I take things as they come.

CPerf10 I prefer to just let things happen.

OAesA7 I seldom notice the emotional aspects of paintings and pictures.

OAesA8 I do not like poetry.

OAesA10 I do not enjoy watching dance performances.

HMode1 I don't think that I'm better than other people.

HMode6 I believe that I am better than others.

HMode8 I am more capable than most others.

HMode7 I like to attract attention.

HMode9 I am likely to show off if I get the chance.

HMode10 I boast about my virtues.

HFair5 I try to follow the rules.

OUnc04 I rebel against authority.

OUnc05 I swim against the current.

OInqu1 I am interested in science.

OInqu3 I enjoy intellectual games.

CPerf1 I pay attention to details.

CPerf3 I have an eye for detail.

CPerf9 I pay too little attention to details.

CPrud8 I make careless mistakes.

CPrud10 I make a fool of myself.

HSinc2 I use flattery to get ahead.

HSinc3 I tell other people what they want to hear so that they will do what I want them to do.

OInqu5 I find political discussions interesting.

OInqu9 I don't bother worrying about political and social problems.

EFear1 I like to do frightening things.

EFear8 I love dangerous situations.

HGree7 I wish to stay young forever.

	[Factor_40]	[Factor_20]	[Factor_39]	[Factor_40]	[Factor_39]	[Factor_20]	[Factor_40]	[Factor_22]	[Factor_21]	[Factor_40]	[Factor_40]
XSocB10											
AForg1	[Factor_41]	[Factor_42]	[Factor_48]	[Factor_42]	[Factor_47]	[Factor_41]	[Factor_41]	[Factor_42]	[Factor_47]	[Factor_41]	[Factor_41]
AForg4	[Factor_41]	[Factor_42]	[Factor_3]	[Factor_42]	[Factor_3]	[Factor_41]	[Factor_41]	[Factor_42]	[Factor_3]	[Factor_41]	[Factor_41]
HSinc10	[Factor_41]	[Factor_42]	[Factor_3]	[Factor_42]	[Factor_3]	[Factor_41]	[Factor_41]	[Factor_42]	[Factor_3]	[Factor_41]	[Factor_41]
HFair6	[Factor_42]	[Factor_43]	[Factor_17]	[Factor_43]	[Factor_41]	[Factor_11]	[Factor_14]	[Factor_15]	[Factor_14]	[Factor_15]	[Factor_13]
HFair7	[Factor_42]	[Factor_43]	[Factor_17]	[Factor_43]	[Factor_41]	[Factor_26]	[Factor_14]	[Factor_15]	[Factor_14]	[Factor_15]	[Factor_13]
HFair9	[Factor_42]	[Factor_43]	[Factor_17]	[Factor_43]	[Factor_41]	[Factor_26]	[Factor_14]	[Factor_15]	[Factor_14]	[Factor_15]	[Factor_13]
AFlex1	[Factor_43]	[Factor_44]	[Factor_41]	[Factor_44]	[Factor_42]	[Factor_42]	[Factor_43]	[Factor_42]	[Factor_42]	[Factor_42]	[Factor_42]
AGent2	[Factor_43]	[Factor_44]	[Factor_41]	[Factor_44]	[Factor_42]	[Factor_42]	[Factor_43]	[Factor_42]	[Factor_42]	[Factor_42]	[Factor_42]
CPerf10	[Factor_43]	[Factor_44]	[Factor_41]	[Factor_44]	[Factor_42]	[Factor_42]	[Factor_43]	[Factor_42]	[Factor_42]	[Factor_42]	[Factor_42]
OAesA7	[Factor_44]	[Factor_45]	[Factor_42]	[Factor_45]	[Factor_43]	[Factor_43]	[Factor_44]	[Factor_41]	[Factor_43]	[Factor_43]	[Factor_43]
OAesA8	[Factor_44]	[Factor_45]	[Factor_42]	[Factor_45]	[Factor_43]	[Factor_43]	[Factor_44]	[Factor_41]	[Factor_43]	[Factor_43]	[Factor_43]
OAesA10	[Factor_44]	[Factor_45]	[Factor_42]	[Factor_45]	[Factor_43]	[Factor_43]	[Factor_44]	[Factor_41]	[Factor_43]	[Factor_43]	[Factor_43]
HMode1	[Factor_45]	[Factor_7]	[Factor_45]	[Factor_9]	[Factor_10]	[Factor_9]	[Factor_8]	[Factor_46]	[Factor_9]	[Factor_45]	[Factor_46]
HMode6	[Factor_45]	[Factor_7]	[Factor_45]	[Factor_9]	[Factor_10]	[Factor_9]	[Factor_8]	[Factor_46]	[Factor_9]	[Factor_45]	[Factor_46]
HMode8	[Factor_45]	[Factor_7]	[Factor_45]	[Factor_9]	[Factor_10]	[Factor_9]	[Factor_8]	[Factor_46]	[Factor_9]	[Factor_45]	[Factor_46]
HMode7	[Factor_46]	[Factor_47]	[Factor_44]	[Factor_47]	[Factor_45]	[Factor_46]	[Factor_46]	[Factor_47]	[Factor_45]	[Factor_46]	[Factor_45]
HMode9	[Factor_46]	[Factor_47]	[Factor_44]	[Factor_47]	[Factor_45]	[Factor_46]	[Factor_46]	[Factor_47]	[Factor_45]	[Factor_46]	[Factor_45]
HMode10	[Factor_46]	[Factor_47]	[Factor_44]	[Factor_47]	[Factor_45]	[Factor_46]	[Factor_46]	[Factor_47]	[Factor_45]	[Factor_46]	[Factor_45]
HFair5	[Factor_47]	[Factor_48]	[Factor_46]	[Factor_48]	[Factor_46]	[Factor_48]	[Factor_47]	[Factor_48]	[Factor_46]	[Factor_47]	[Factor_47]
OUnc04	[Factor_47]	[Factor_48]	[Factor_46]	[Factor_48]	[Factor_46]	[Factor_48]	[Factor_47]	[Factor_46]	[Factor_47]	[Factor_47]	[Factor_47]
OUnc05	[Factor_47]	[Factor_48]	[Factor_46]	[Factor_48]	[Factor_46]	[Factor_48]	[Factor_48]	[Factor_46]	[Factor_47]	[Factor_46]	[Factor_47]
OInqu1	[Factor_48]	[Factor_11]	[Factor_14]	[Factor_48]	[Factor_12]	[Factor_48]	[Factor_13]	[Factor_48]	[Factor_48]	[Factor_48]	[Factor_48]
OInqu3	[Factor_48]	[Factor_11]	[Factor_14]	[Factor_48]	[Factor_12]	[Factor_48]	[Factor_13]	[Factor_48]	[Factor_48]	[Factor_48]	[Factor_48]
CPerf1	[Factor_49]										
CPerf3	[Factor_49]										
CPerf9	[Factor_49]										
CPrud8	[Factor_50]	[Factor_50]	[Factor_50]	[Factor_50]	[Factor_50]	[Factor_51]	[Factor_50]	[Factor_50]	[Factor_50]	[Factor_50]	[Factor_8]
CPrud10	[Factor_50]	[Factor_8]									
HSinc2	[Factor_51]	[Factor_52]	[Factor_17]	[Factor_51]	[Factor_51]	[Factor_26]	[Factor_14]	[Factor_15]	[Factor_14]	[Factor_15]	[Factor_13]
HSinc3	[Factor_51]	[Factor_52]	[Factor_17]	[Factor_51]	[Factor_51]	[Factor_26]	[Factor_14]	[Factor_15]	[Factor_14]	[Factor_15]	[Factor_13]
OInqu5	[Factor_52]	[Factor_22]	[Factor_52]	[Factor_23]	[Factor_52]	[Factor_53]	[Factor_52]	[Factor_24]	[Factor_52]	[Factor_51]	[Factor_52]
OInqu9	[Factor_52]	[Factor_22]	[Factor_52]	[Factor_23]	[Factor_52]	[Factor_53]	[Factor_52]	[Factor_24]	[Factor_52]	[Factor_51]	[Factor_52]
EFear1	[Factor_53]	[Factor_53]	[Factor_52]	[Factor_53]	[Factor_54]	[Factor_53]	[Factor_53]	[Factor_51]	[Factor_53]	[Factor_52]	[Factor_53]
EFear8	[Factor_53]	[Factor_53]	[Factor_52]	[Factor_53]	[Factor_54]	[Factor_53]	[Factor_53]	[Factor_51]	[Factor_53]	[Factor_52]	[Factor_53]
HGree7	[Factor_54]	[Factor_54]	[Factor_54]	[Factor_53]	[Factor_54]	[Factor_0]	[Factor_54]	[Factor_52]	[Factor_54]	[Factor_53]	[Factor_54]
Fit Score						94.5946%	91.0811%	95.6888%	95.4054%	90.2453%	94.5946%
						94.5946%	91.0811%	95.6888%	95.4054%	90.2453%	94.5946%

Close Save As... Print Clustering Frequency Graph

Associated graph 1.xlsb

AGent8 ①

lex5 ①

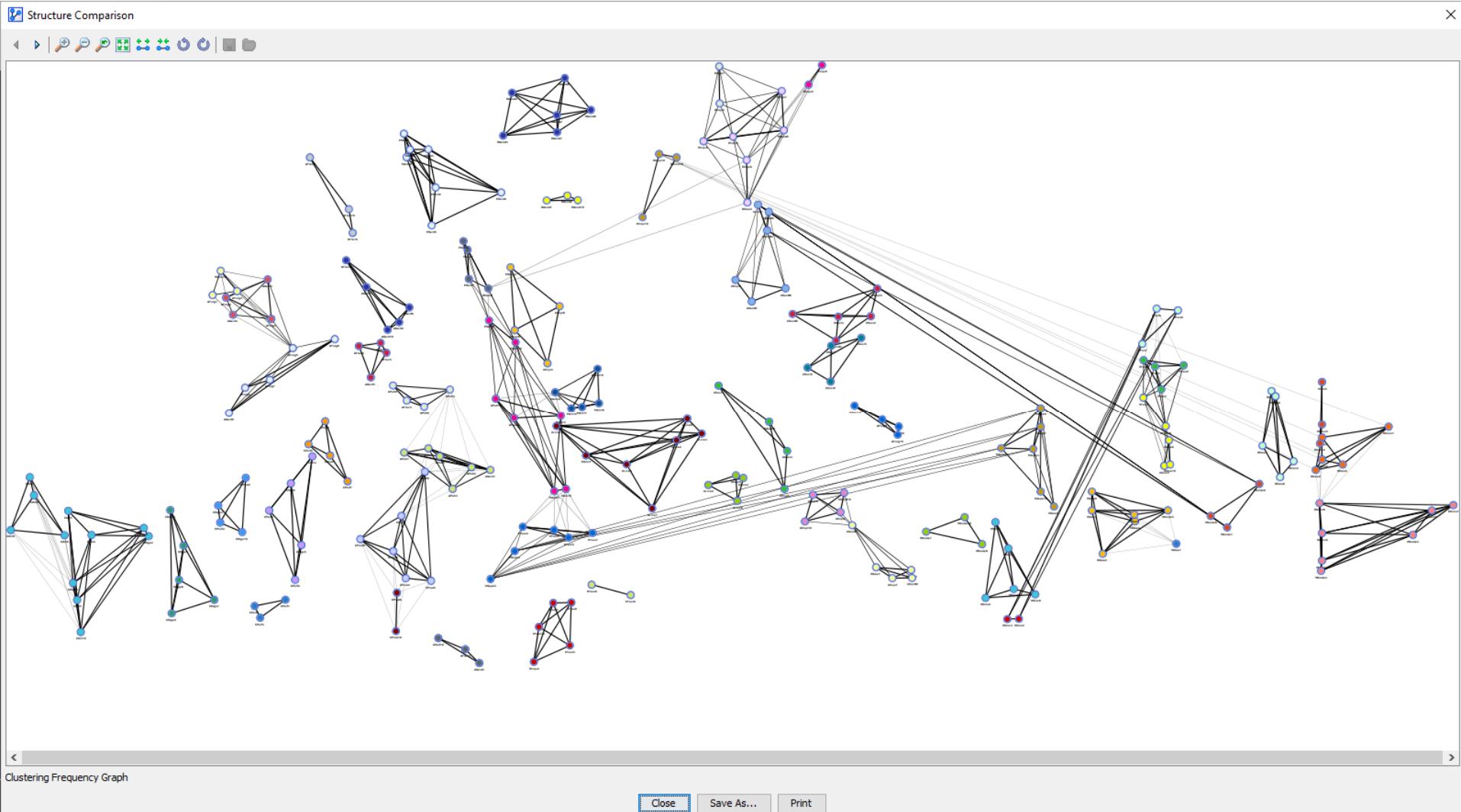
APati6 ①

AFlex3 ①

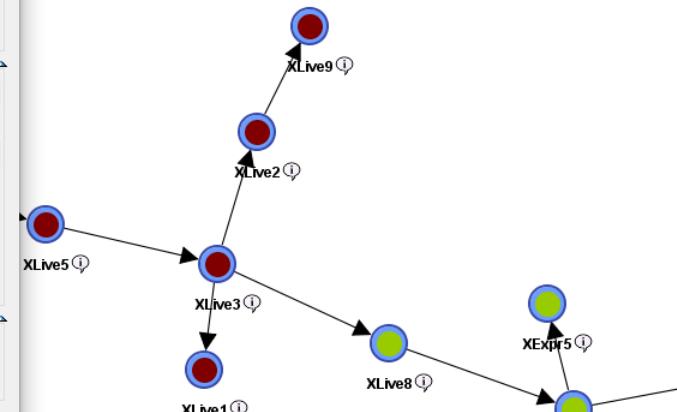
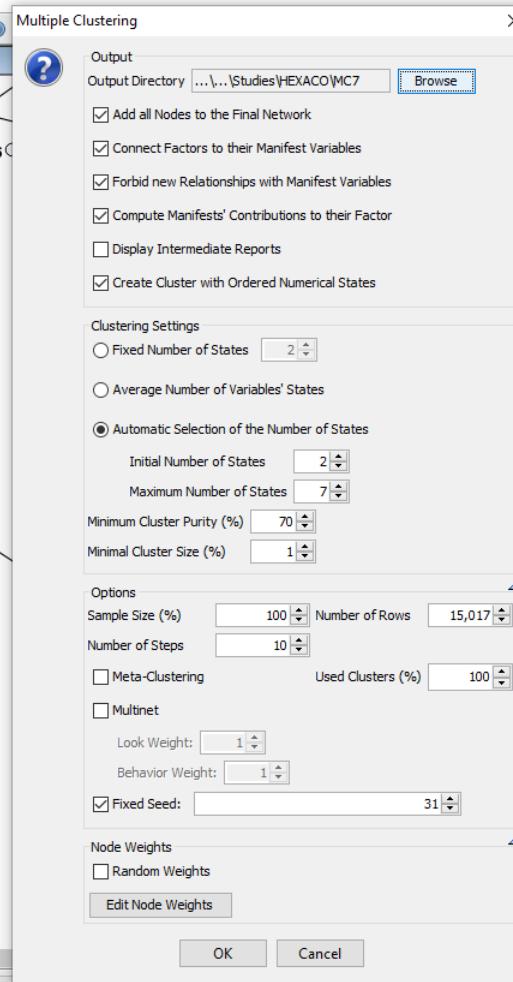
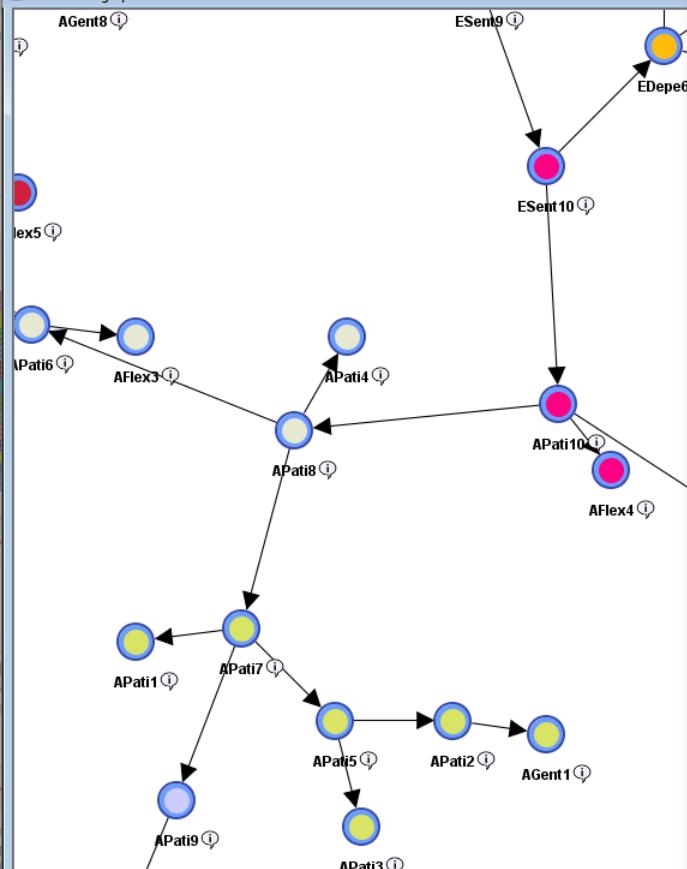
APati1 ①

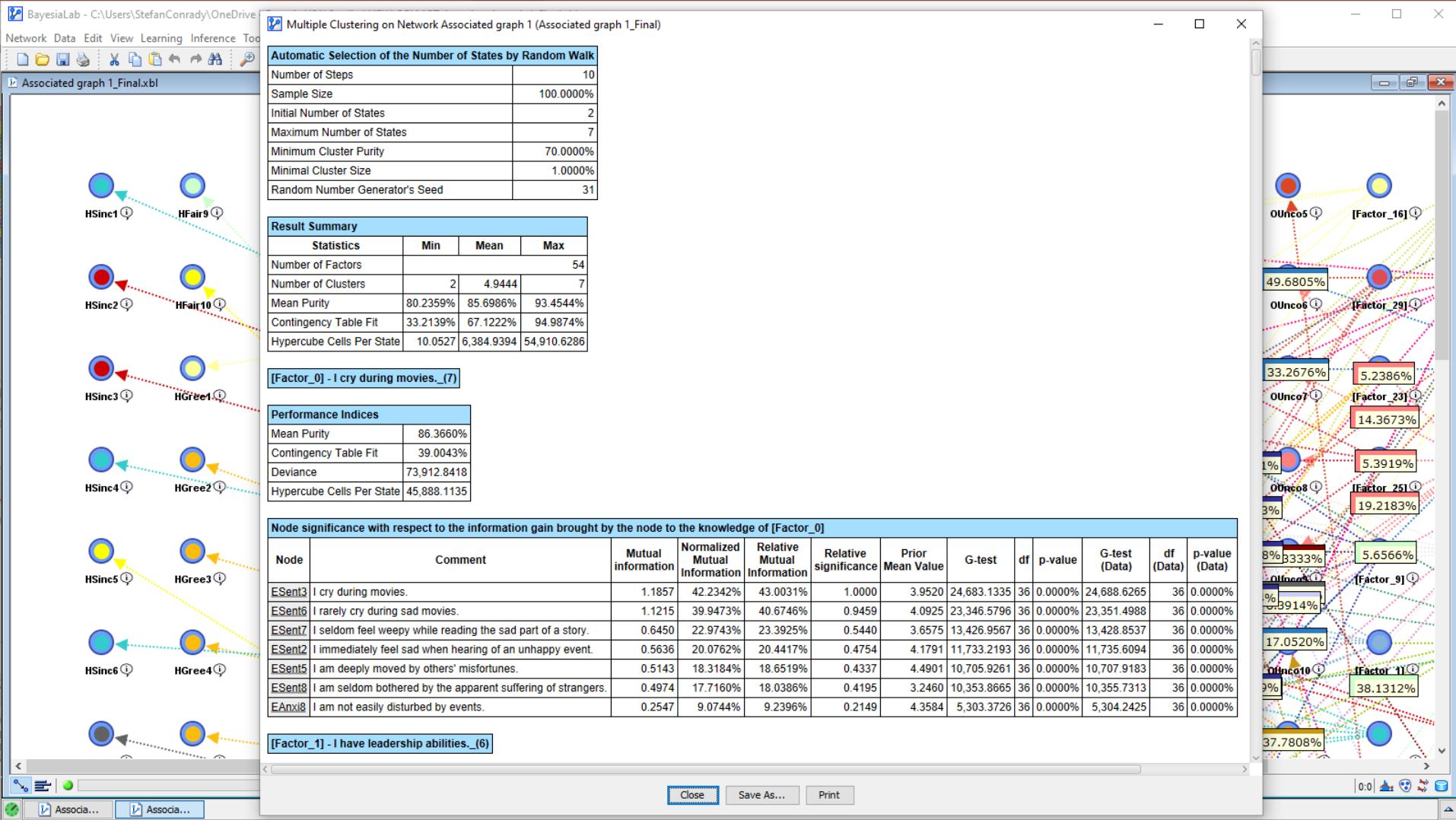
APati9 ①

Associa...



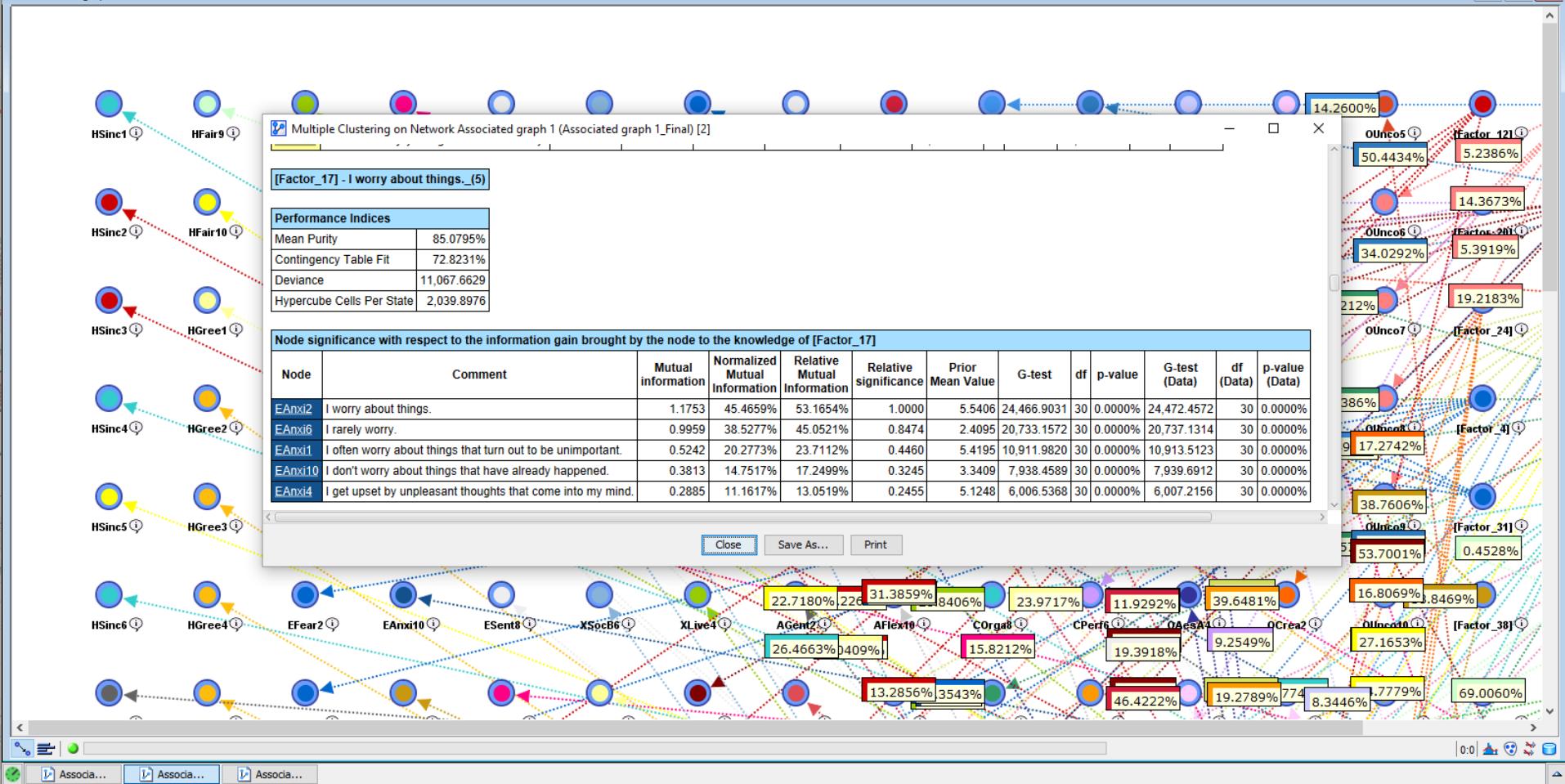
Associated graph 1.xaml *





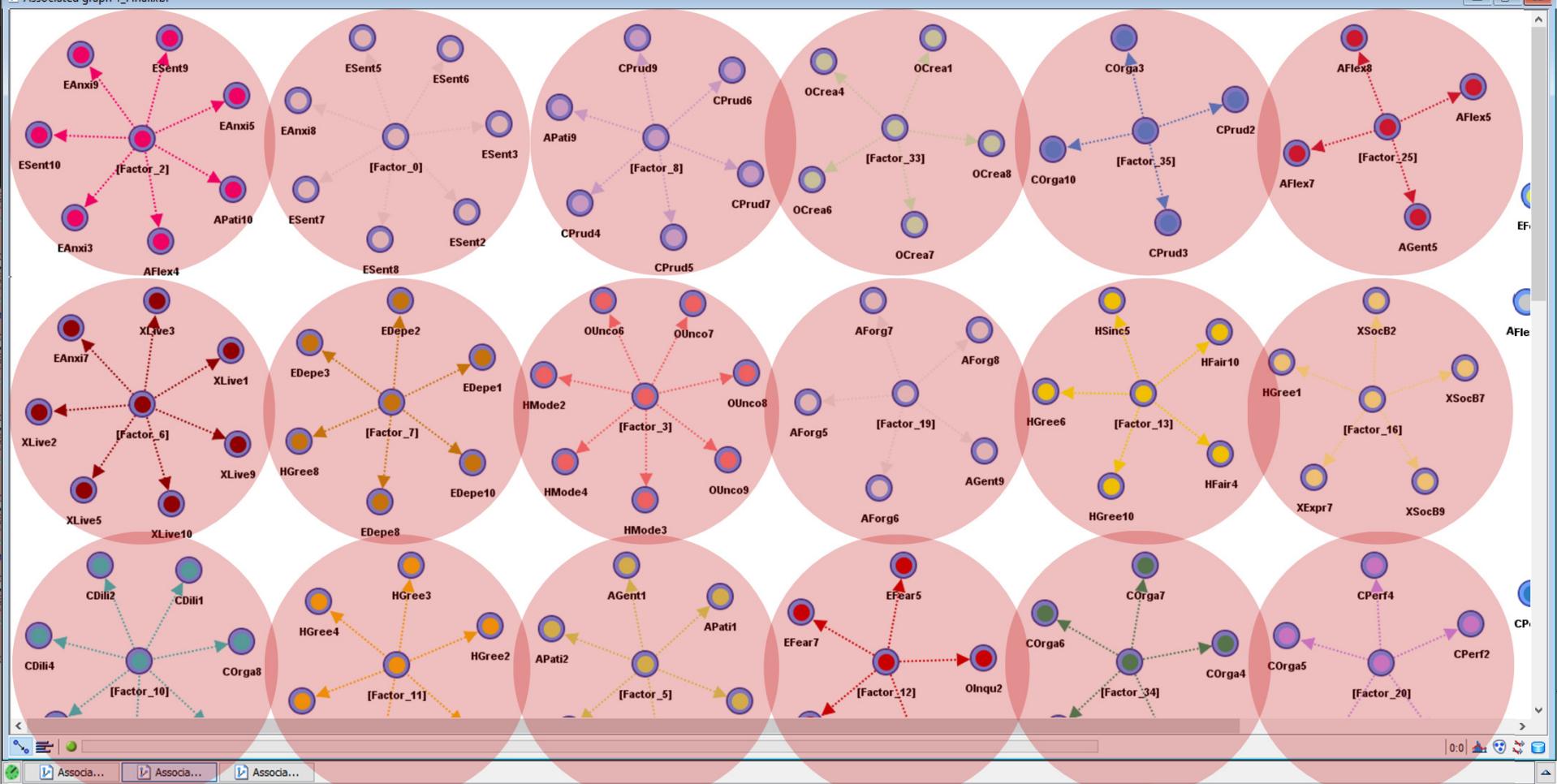


Associated graph 1_Final.xbl



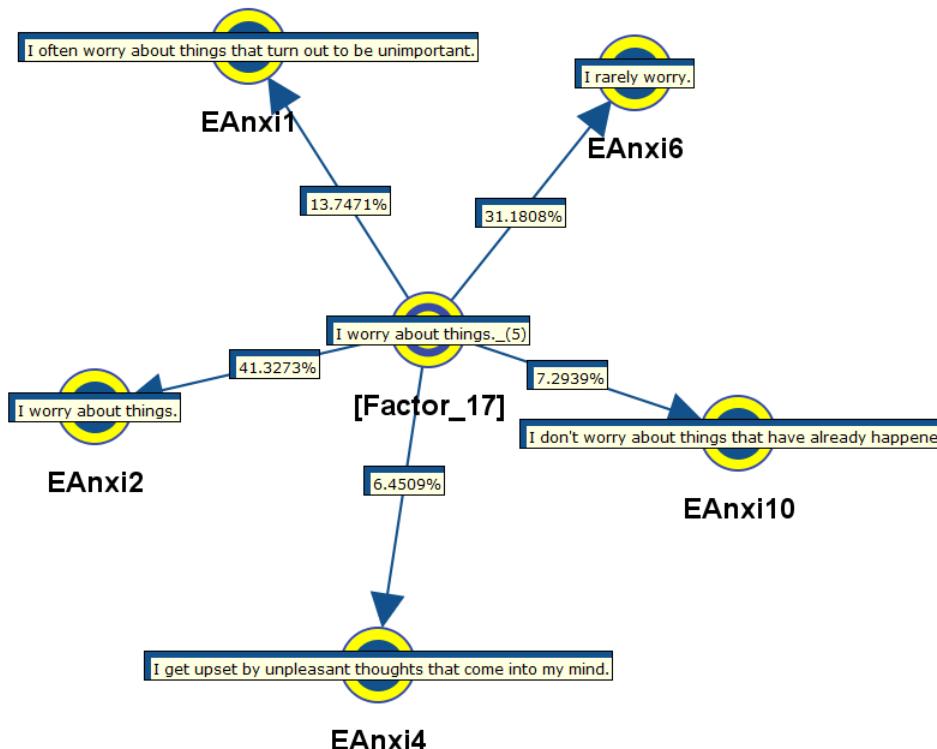


Associated graph 1_Final.xbl*



Newly-generated states of Factor 17

Numerical values of the factor states are the weighted averages of the values of the manifest states



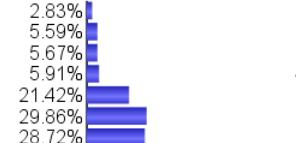
0.00%

value: 4,366

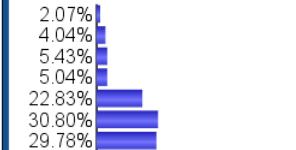
[Factor_17]
Mean: 4.360 Dev: 0.199
Value: 4.360



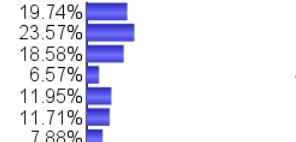
EAnxi1
Mean: 5.419 Dev: 1.585
Value: 5.419



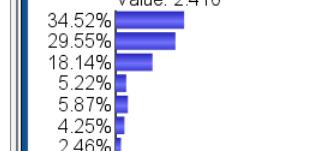
EAnxi2
Mean: 5.541 Dev: 1.473
Value: 5.541



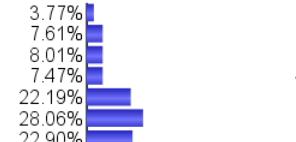
EAnxi10
Mean: 3.341 Dev: 1.941
Value: 3.341



EAnxi6
Mean: 2.410 Dev: 1.546
Value: 2.410



EAnxi4
Mean: 5.125 Dev: 1.692
Value: 5.125



Associated graph 1_[Factor_17].xbel * Joint Probability: 100.00%

Target Mean Analysis

Node: X: Y:

I worry about things_(5) Mean

Variable Means	I worry about things	I often worry about things that turn out to be unimportant	I get upset by unpleasant thoughts that come into my mind	I don't worry about things that have already happened	I rarely worry	I worry about things_(5)
1.0	4.15	3.95	4.00	4.50	4.50	4.50
2.0	4.25	4.45	4.12	4.45	4.45	4.45
3.0	4.28	4.35	4.15	4.25	4.25	4.25
4.0	4.28	4.08	4.18	4.25	4.10	4.25
5.0	4.32	4.25	4.32	4.25	4.05	4.25
6.0	4.42	4.55	4.45	4.25	3.95	4.45
7.0	4.48	4.22	4.48	4.22	3.90	4.48

Variables

- All Curves
- I worry about things.
- I often worry about things that turn out to be unimportant.
- I get upset by unpleasant thoughts that come into my mind.
- I don't worry about things that have already happened.
- I rarely worry.

Close Save

0.0

Associated graph 1_[Factor_17].xbel Associated graph 1_[Factor_17].xbel Associated graph 1_[Factor_17].xbel

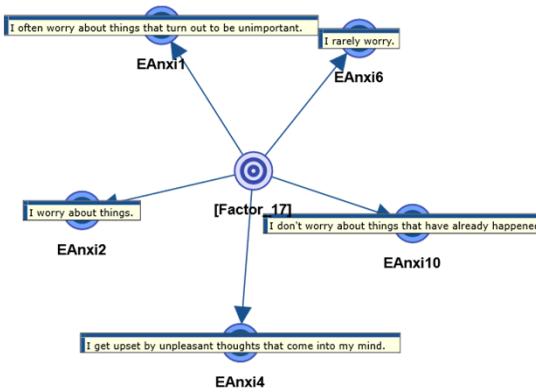
Quality of Representation

Key Measures

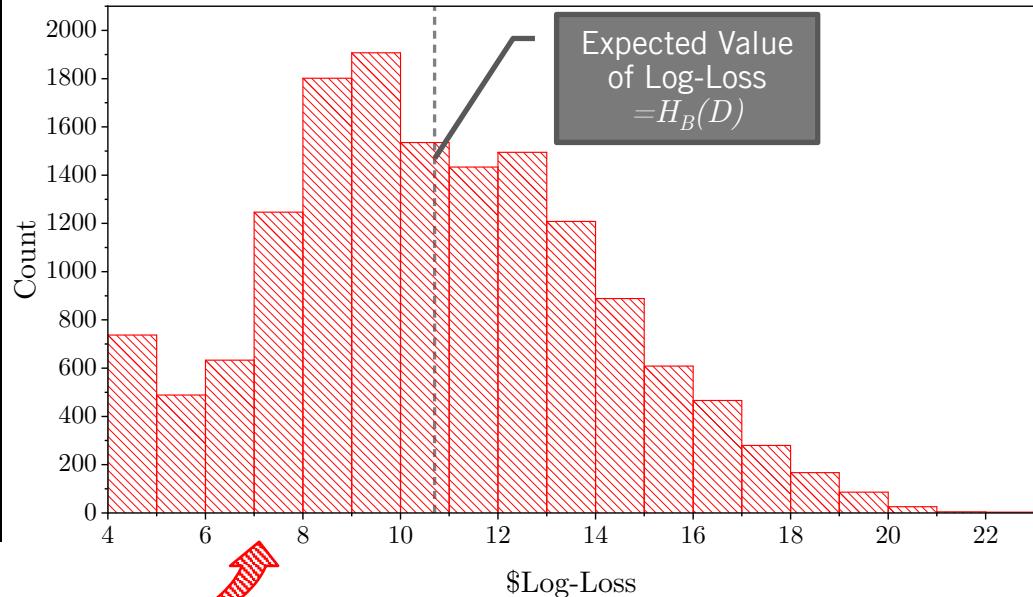
- Contingency Table Fit
 - Log-Loss Distribution
 - Entropy
- Purity
- Deviance
- Hypercube Cells Per State

Quality Measures

Log-Loss Distribution of Factor 17



EAnxi1	EAnxi2	EAnxi4	EAnxi6	EAnxi10	\$Log-Loss
5	6	6	2	2	7.45
7	6	6	2	2	7.22
6	6	6	1	2	7.39
7	5	7	2	5	13.16
2	2	4	4	5	16.40
5	5	3	3	3	9.37
5	6	5	1	3	10.66
5	6	6	3	5	10.89
7	7	7	1	1	4.59
5	5	5	3	5	8.58
5	5	2	3	3	9.63
7	5	6	2	2	10.23
7	7	7	1	1	4.59
7	7	7	1	4	8.73
7	7	7	1	1	4.59

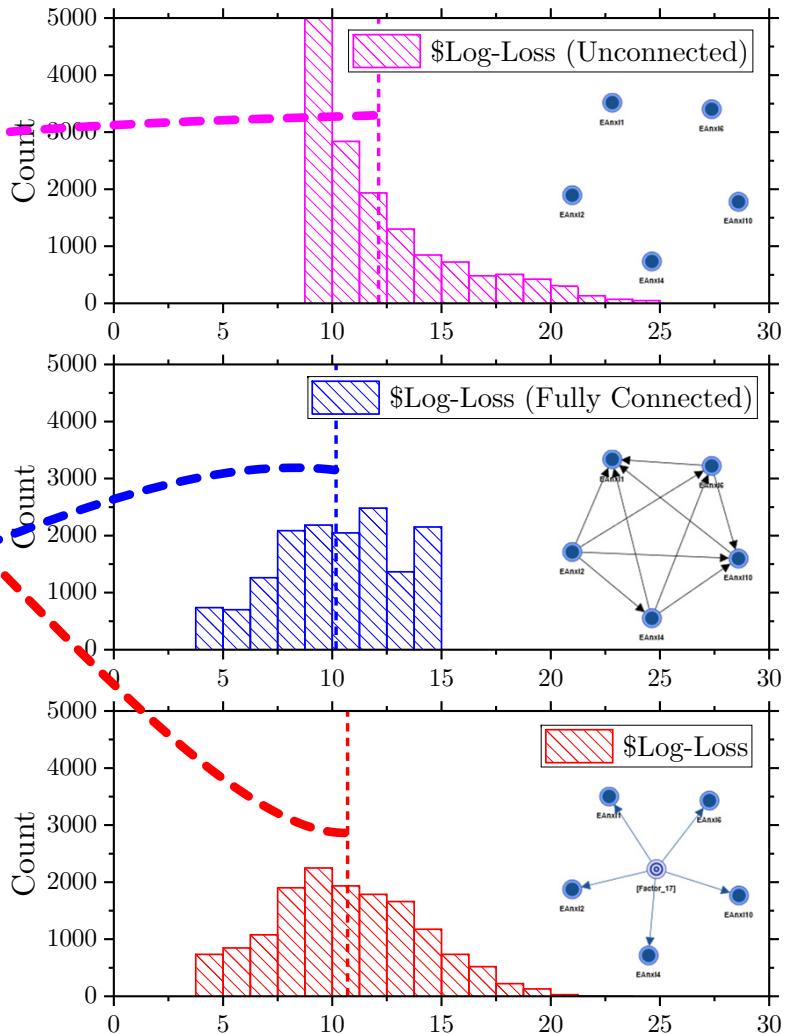


Quality Measures

Contingency Table Fit (CTF)

$$C_B = 100 \times \frac{H_U(D) - H_B(D)}{H_U(D) - H_F(D)}$$

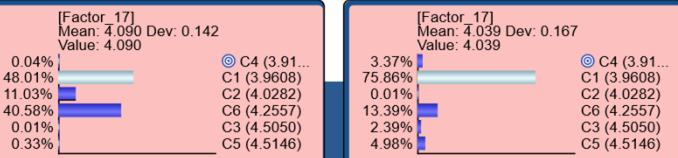
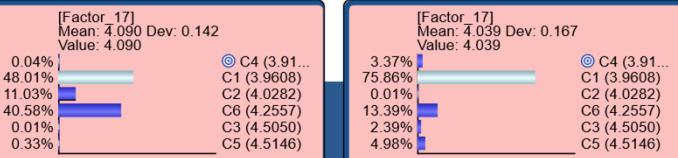
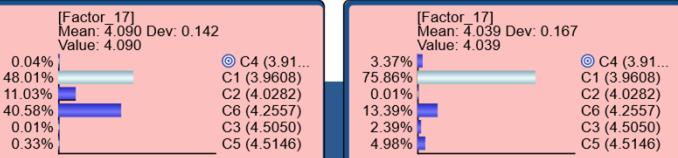
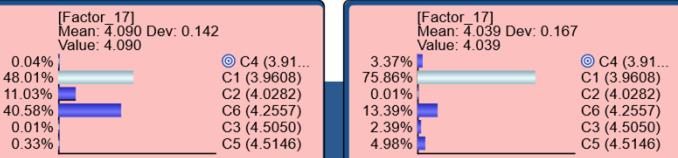
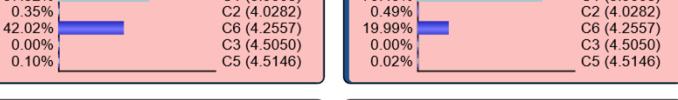
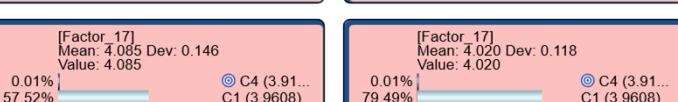
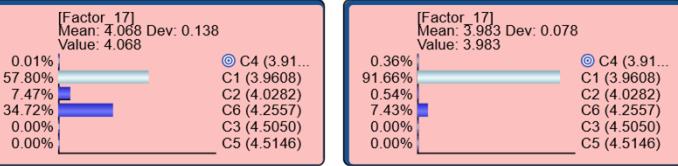
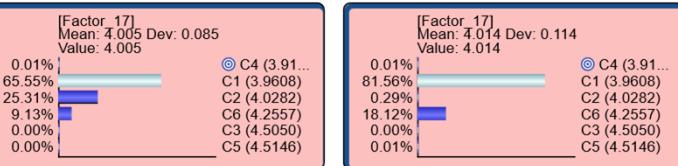
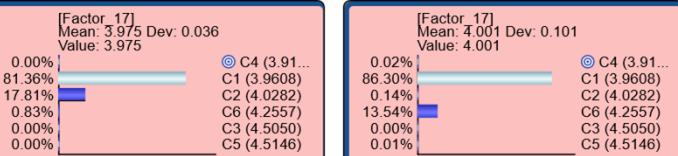
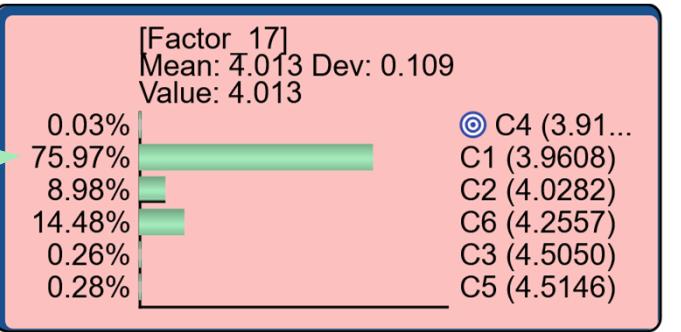
- The CTF measures the quality of the representation of the Joint Probability Distribution by the network.



BayesiaLab Workflow

Purity of State C1

State C1 is
76% “pure”



Associated graph 1_[Factor_17].xbel*

Joint Probability: 100.00%

Network Overall Performance Associated graph 1_[Factor_17].xbel (Associated graph 1_[Factor_17])

Learning Database

Probability

Entropy (H): 10.7012 Standard Deviation: 3.3409
Normalized Entropy (Hn): 76.2368%
Hn(Complete): 72.4493% Hn(Unconnected): 86.3857%
Contingency Table Fit: 72.8231%
Deviance: 11,067.6629
Number of Processed Observations: 15,017

Density Function Number of Intervals

Distribution Function

Close Save As... Print Extract Database

EAnxi1
Mean: 5.419 Dev: 1.585
Value: 5.419 (+2.771)

(3.911...	2.83%
9608)	5.59%
0282)	5.67%
2557)	5.91%
5050)	21.42%
5146)	29.86%
	28.72%

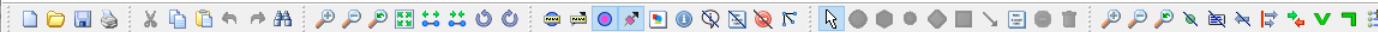
EAnxi10
Mean: 3.341 Dev: 1.941
Value: 3.341 (-2.475)

19.74%	1
23.57%	2
18.58%	3
6.57%	4
11.95%	5
11.71%	6
7.88%	7

EAnxi4
Mean: 5.125 Dev: 1.692
Value: 5.125 (+1.734)

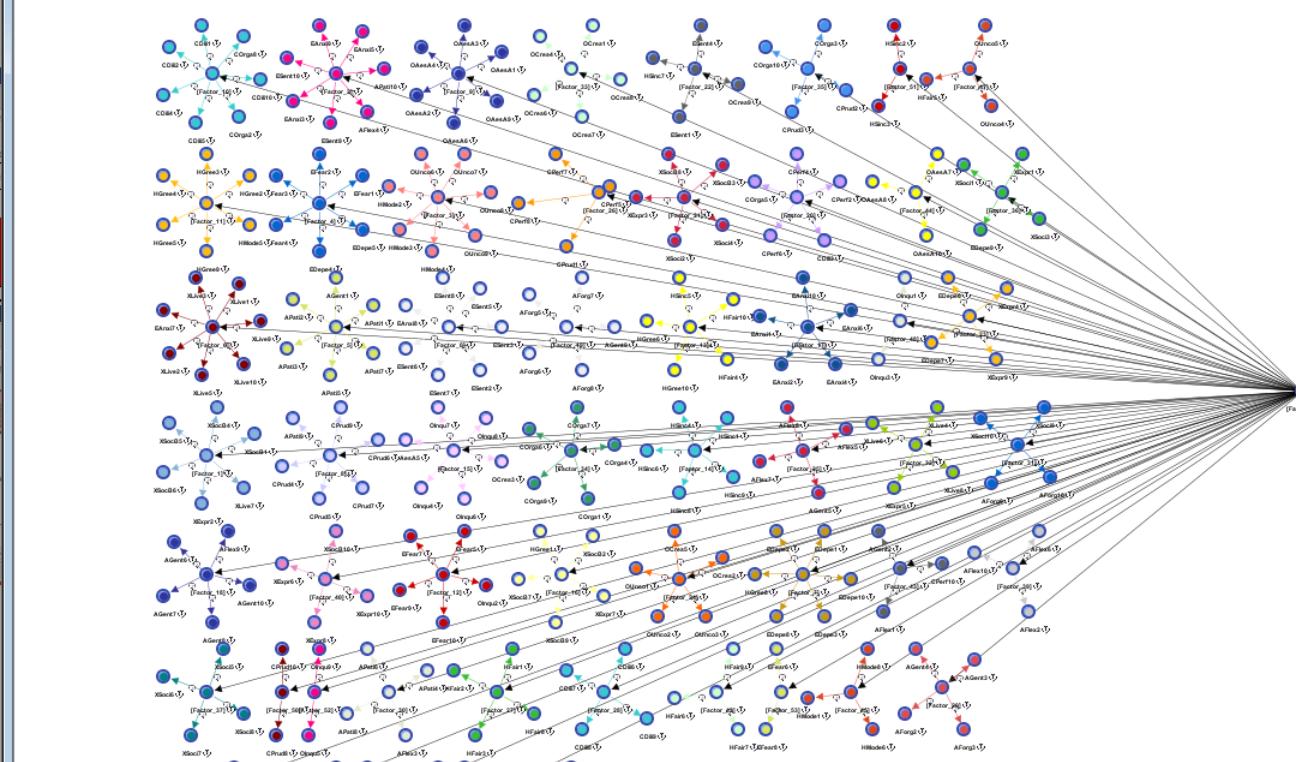
3.77%	1
7.61%	2
8.01%	3
7.47%	4
22.19%	5
28.06%	6
22.90%	7

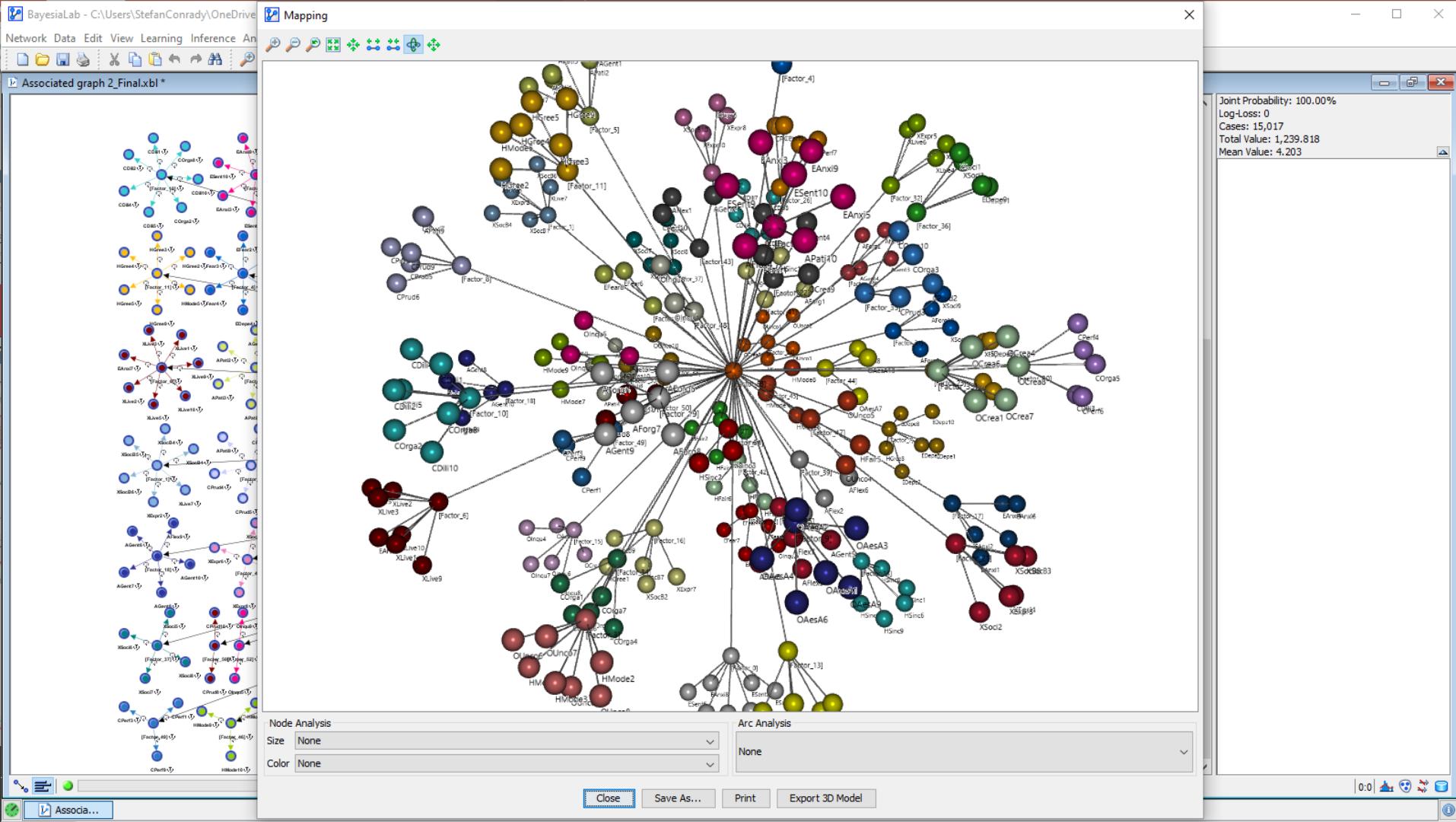
Network Data Edit View Learning Inference Analysis Monitor Tools Window Help



Associated graph 2_Final.xbl *

Joint Probability: 100.00%
Log-Loss: 0
Cases: 15,017
Total Value: 1,239.818
Mean Value: 4.203







BAYESIALAB

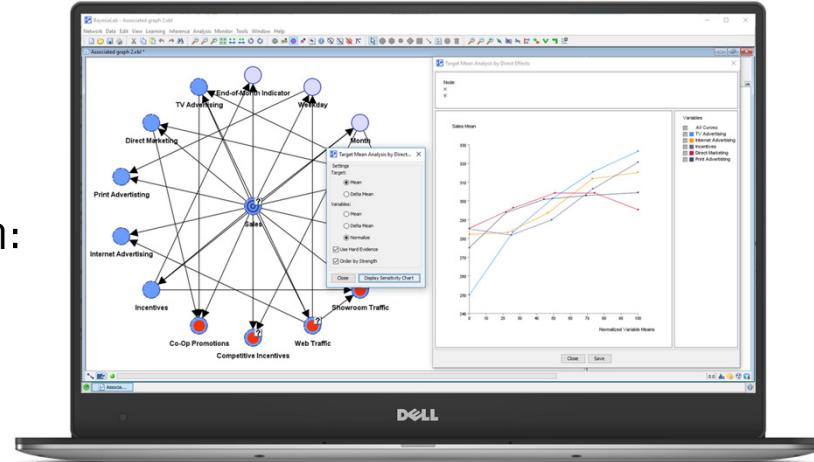
VR

In Conclusion...

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New Delhi, India
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Chicago, IL
- December 4–6
New York, NY



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Thank You!



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