



Reasoning About Renewable Energy

Overcoming the Flaw of Averages with Bayesian Networks

May 14, 2020

Agenda

Introduction & Resources

Motivation & Objective

- Sun & wind: unlimited, free power?
- Reasoning under extreme uncertainty

Analysis Workflow

- Machine-learning a Bayesian network from weather data
- Building a Bayesian network model from domain knowledge
- Simulating and evaluating scenarios

Introduction



STEFAN CONRADY
Managing Partner

BAYESIA USA, LLC

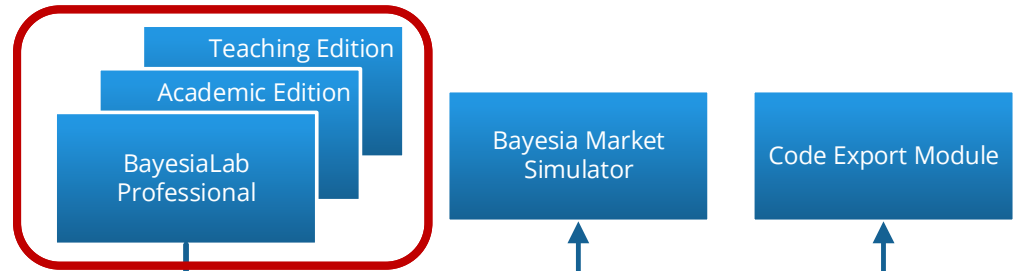
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Franklin, TN 37069
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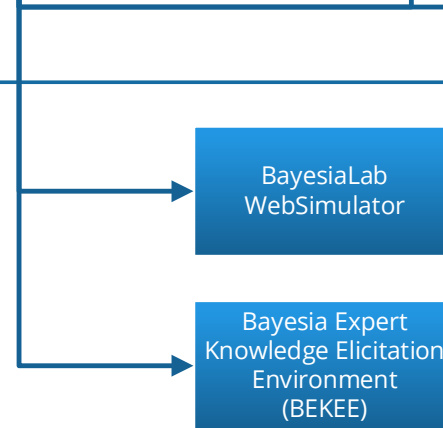
BAYESIALAB



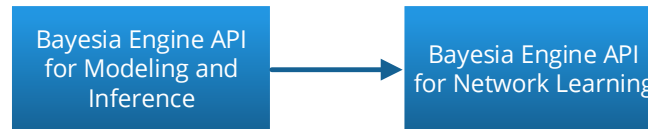
Desktop
Software



Web
Application



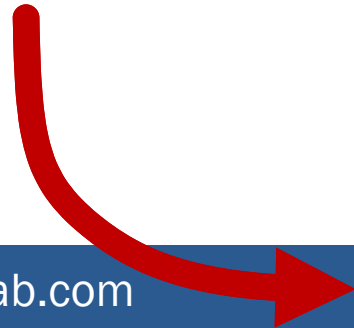
API



Resources

Webinar Materials Available in the BayesiaLab Community

- Slides
- Network Model
- Webinar Recording
- Q&A



🔍 Search the community

Question?
"Start a new topic"

Home > Discussions

COVID-19 Diagnostics with Bayesian Networks

Follow

Active Newest Popular



Call for Expertise — Help Improve the COVID-19 Knowledge Base

👍 2

Please join our volunteer group of infectious disease specialists, clinicians, epidemiologists, biost...

Stefan Conrady · BAYESIALAB · 4 days ago · 2 replies · 101 · 2 · Last reply 3 days ago

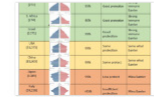


Oral polio vaccine as infection control?

👍 2

Chinese study theorizes that oral polio vaccine can reduce probably of Covid19 in under-30 dem...

Kurt Schulzke · 4 days ago · 2 replies · 5 · 2 · Last reply 4 days ago



Start a new topic

This is a group for researchers who currently work on COVID-19.

CATEGORIES

- Blog
- Events
- > Videos
- > Discussions
 - COVID-19 Diagnostics with Bayesian Networks
 - General Discussions
- > Knowledge Base
- Frequently Asked Questions
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Please see Part 1-4 of this webinar series

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Reasoning Under Uncertainty (Part 2) – Differential Diagnosis of COVID-19

Updated 7 days ago



CONTENTS

- COVID-19 WebSimulator
- From Local Insight to Worldwide Diagnostic Practice
- Call for Expertise
- Overcoming Human Challenges in Reasoning

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SHARE

U.S. Approves Giant Solar Project in Nevada

Gemini project will have capacity to power all Las Vegas homes



The Lake Las Vegas community of Henderson. The project power to cover the residential population of Las Vegas. CF FOR THE WALL STREET JOURNAL

By Timothy Puko
May 11, 2020, 1:00 p.m. EDT

WASHINGTON—The Interior Department has approved for what it says will be the largest solar project in the U.S., a \$1 billion installation in Nevada.

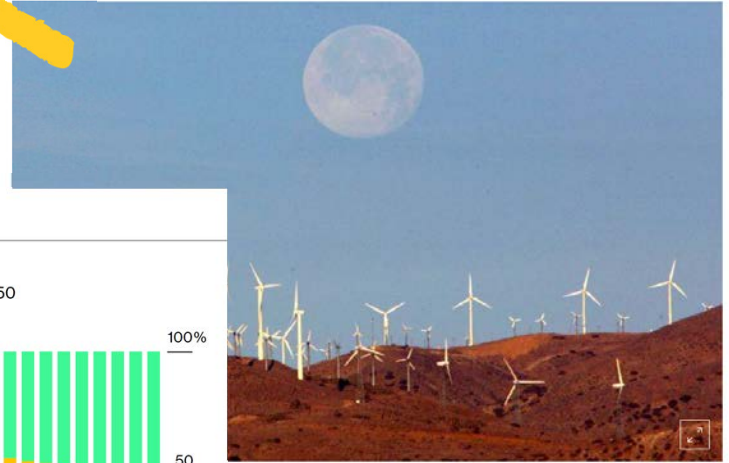


ENVIRONMENT MAY 4, 2020 / 5:31 PM / 9 DAYS AGO

Amid pandemic, U.S. renewable power sources have topped coal for 40 days

Nichola Groom

3 MIN READ



Wind farm in the Mojave Desert in California, January 6,

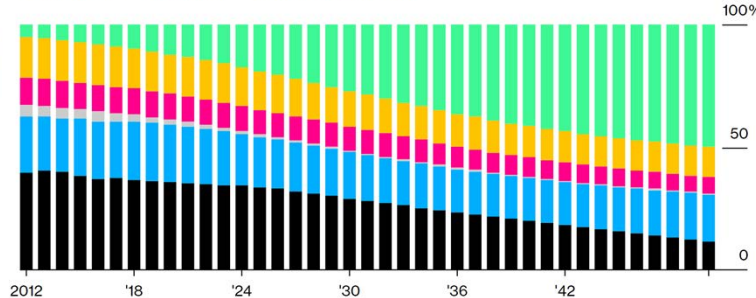
city generated by renewable sources like solar, wind and hydro has topped coal power in the United States for a record 40 straight days, according to U.S. government data released on Monday.

Bloomberg

Power Shift

Wind, solar and other renewables will account for half of all power by 2050

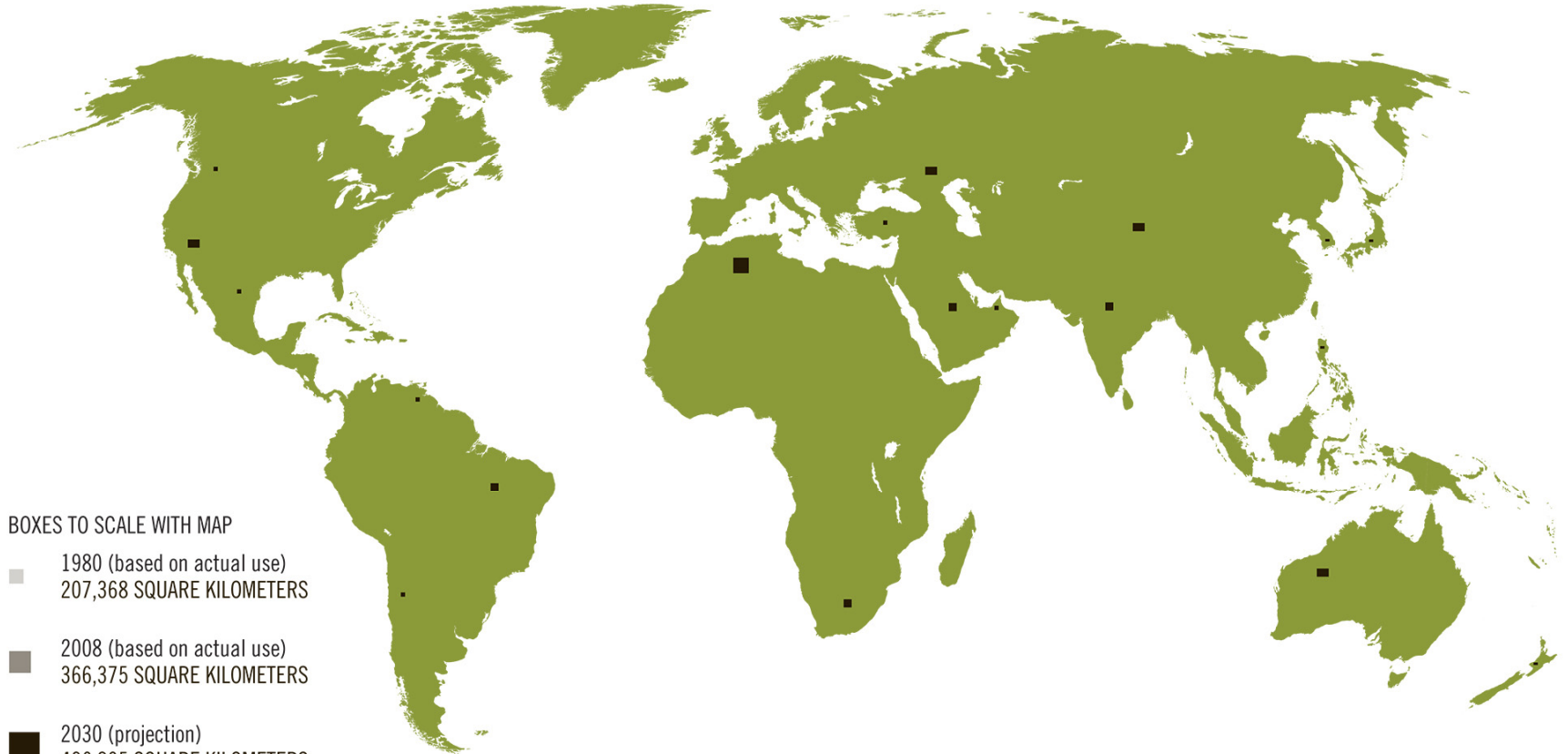
■ Coal ■ Gas ■ Oil ■ Nuclear ■ Hydro ■ Renewables



Source: BloombergNEF

SURFACE AREA REQUIRED TO POWER THE WORLD WITH ZERO CARBON EMISSIONS AND WITH SOLAR ALONE

www.landartgenerator.org



BOXES TO SCALE WITH MAP

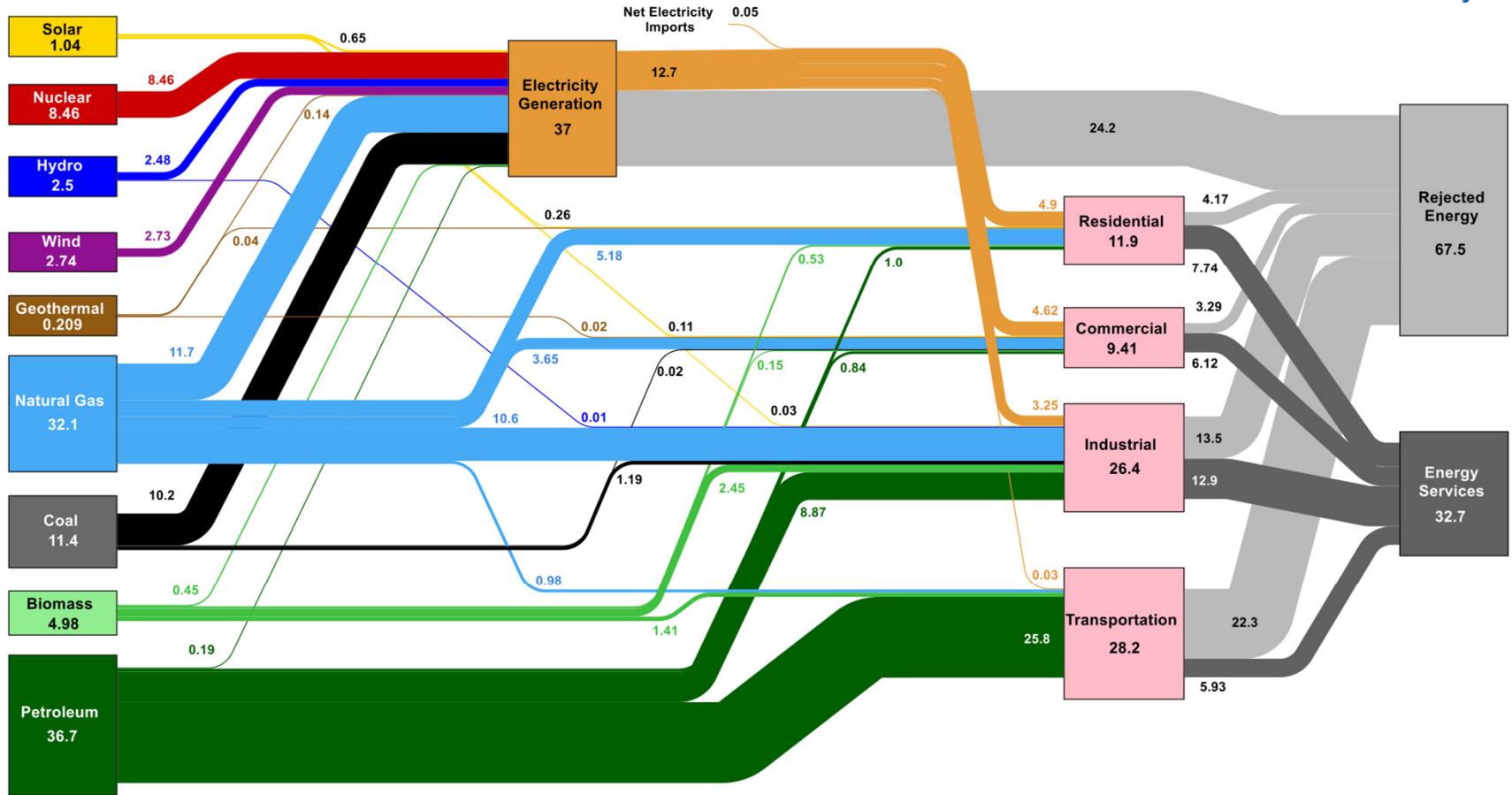
- 1980 (based on actual use)
207,368 SQUARE KILOMETERS
- 2008 (based on actual use)
366,375 SQUARE KILOMETERS
- 2030 (projection)
496,805 SQUARE KILOMETERS

ROOFTOP SOLAR SHIFTS POWER!

LOCAL ENERGY! DISTRIBUTED GENERATION! STRONGER COMMUNITIES!



Estimated U.S. Energy Consumption in 2019: 100.2 Quads



Source: LLNL March, 2020. Data is based on DOE/EIA MER (2019). If this information or a reproduction of it is used, credit must be given to the Lawrence Livermore National Laboratory and the Department of Energy, under whose auspices the work was performed. Distributed electricity represents only retail electricity sales and does not include self-generation. EIA reports consumption of renewable resources (i.e., hydro, wind, geothermal and solar) for electricity in BTU-equivalent values by assuming a typical fossil fuel plant heat rate. The efficiency of electricity production is calculated as the total retail electricity delivered divided by the primary energy input into electricity generation. End use efficiency is estimated as 65% for the residential sector, 65% for the commercial sector, 21% for the transportation sector and 49% for the industrial sector, which was updated in 2017 to reflect DOE's analysis of manufacturing. Totals may not equal sum of components due to independent rounding. LLNL-MI-410527

Objective

Reasoning About Renewable Energy

- Uncertainties regarding future...
 - Energy efficiency
 - Reliability
 - Government interventions
 - Emissions control
 - Manufacturing costs
 - Energy prices
 - Energy demand
 - Consumer behavior

2070
2060
2050
2040
2030

Objective

Typical Questions

- Is it technically feasible to rely on renewables by 2050?
- How much fossil fuels will we still need if we triple the number of solar panels and wind turbines?
- Could batteries provide electricity at night and when there is no wind?
- How big would these batteries have to be?
- What's the risk of blackouts?
- Could we rely exclusively on renewables if electricity usage were cut in half?
- What would it take to shut down all fossil fuel-burning power plants?

Modeling the Problem Domain

Model Requirements

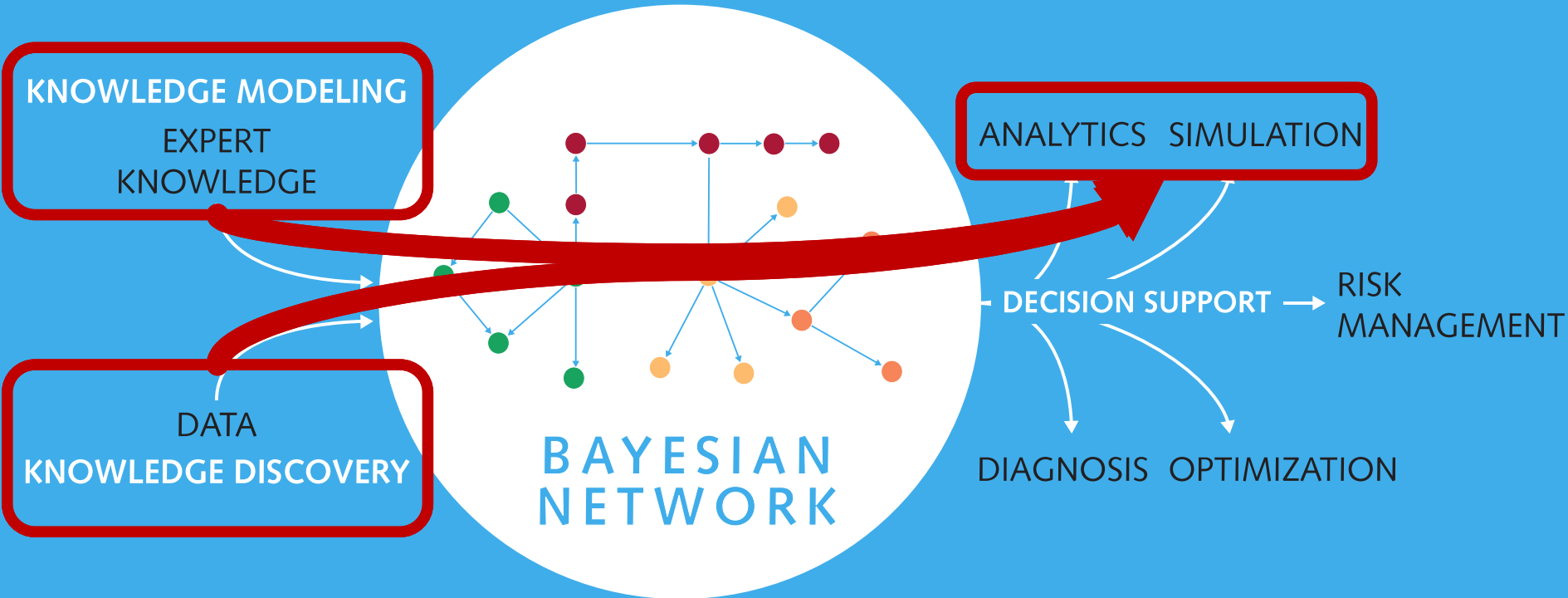
- High level of abstraction, but scalable to include higher level of detail
- High-dimensional
- Probabilistic, to encode uncertainty
- Representing the system, not our reasoning about it
- Causal, to simulate decision/interventions
- Non-static
- Non-linear
- Compatible with expert knowledge and data
- Quick to simulate



SIMULATION
DEPT.



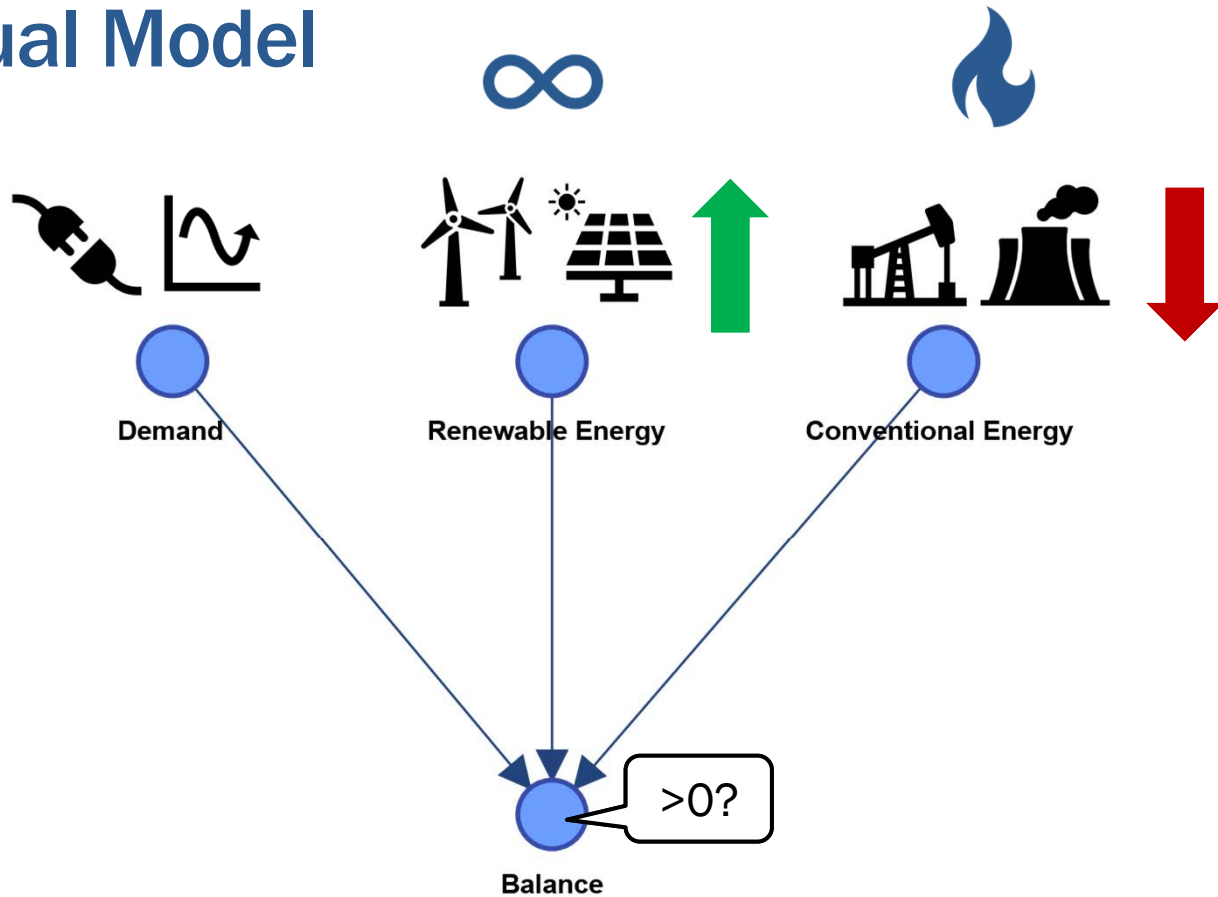
Reasoning Framework: Bayesian Networks



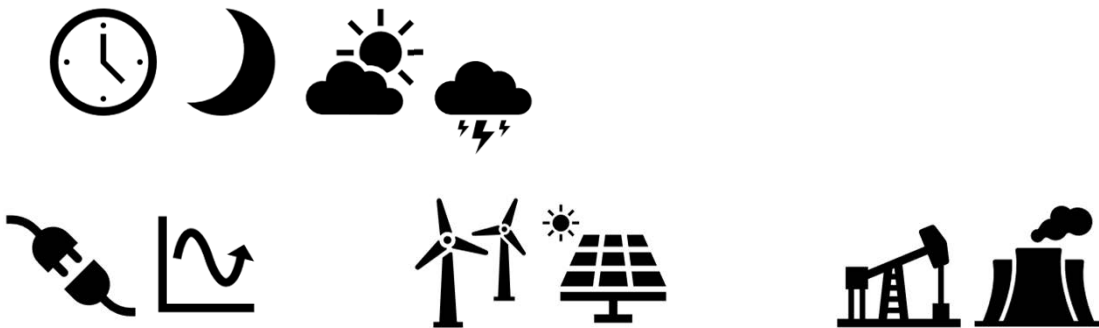
Notes & Caveats

- While we are focusing on energy today, this example is prototypical for any type of “big picture” feasibility analysis.
- The proposed approach is not meant a substitute for detailed system analysis in a later planning process.
- Our model can be a tool for policy makers, investors, analysts, and auditors.
- In this webinar, we do not advocate any particular technology or recommend policies.
- All estimates shown in this webinar are meant to be realistic, but the exact numerical values are arbitrary.

Conceptual Model



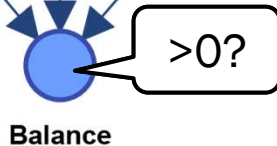




Demand

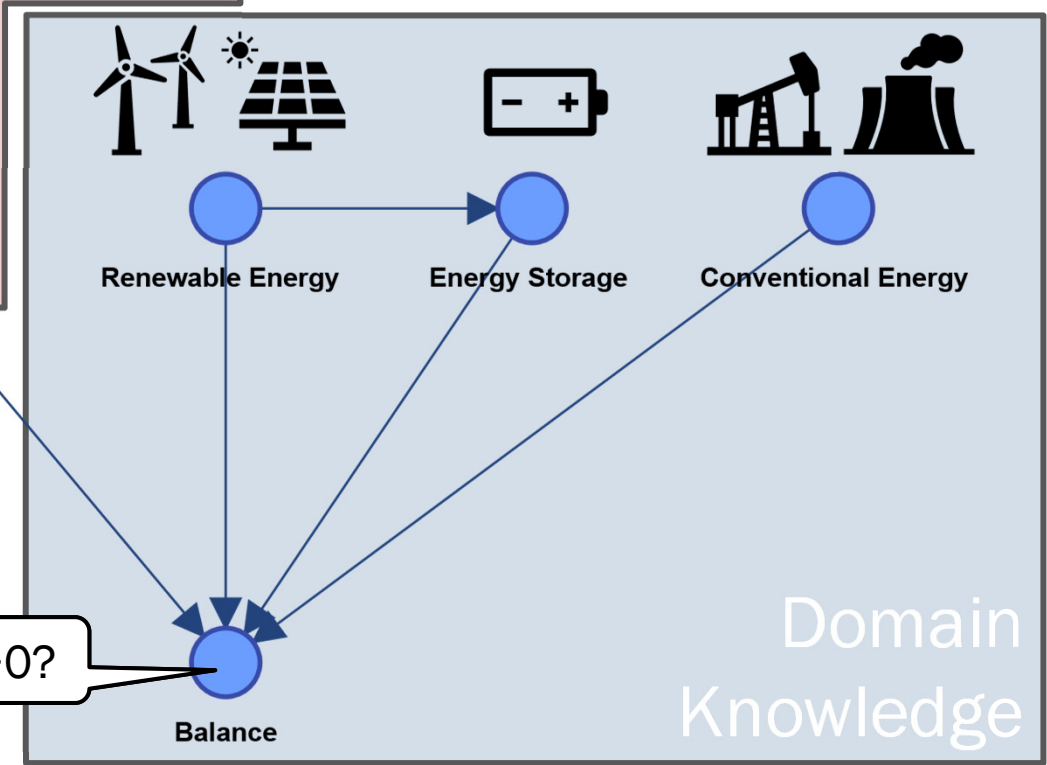
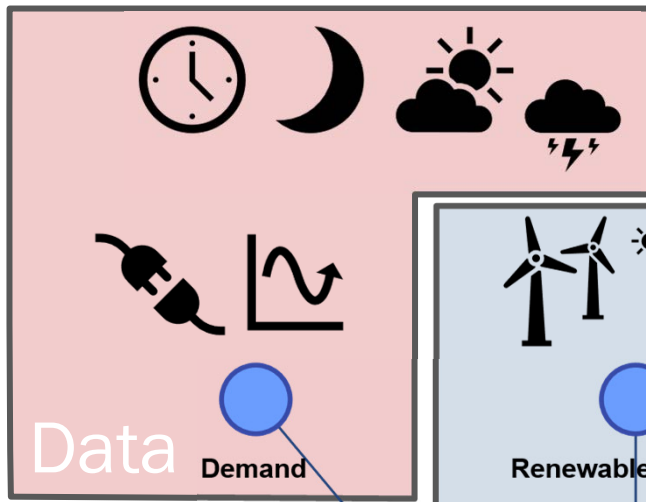
Renewable Energy

Conventional Energy



THE FLAW
OF AVERAGES





Is this possible?

>0?

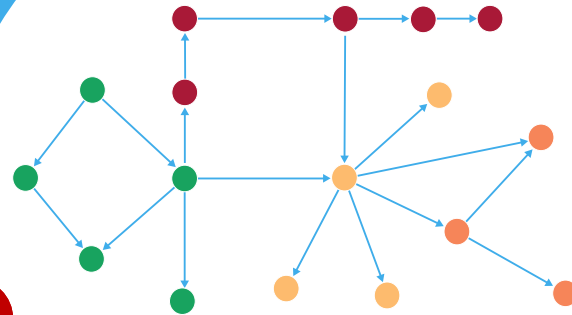
Reasoning Framework: Bayesian Networks

KNOWLEDGE MODELING

EXPERT
KNOWLEDGE

DATA

KNOWLEDGE DISCOVERY



BAYESIAN
NETWORK

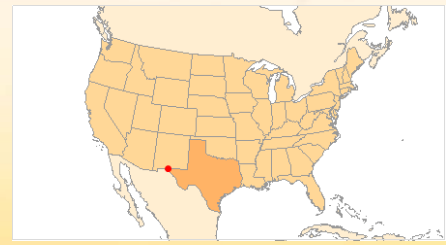
ANALYTICS SIMULATION

DECISION SUPPORT

RISK
MANAGEMENT

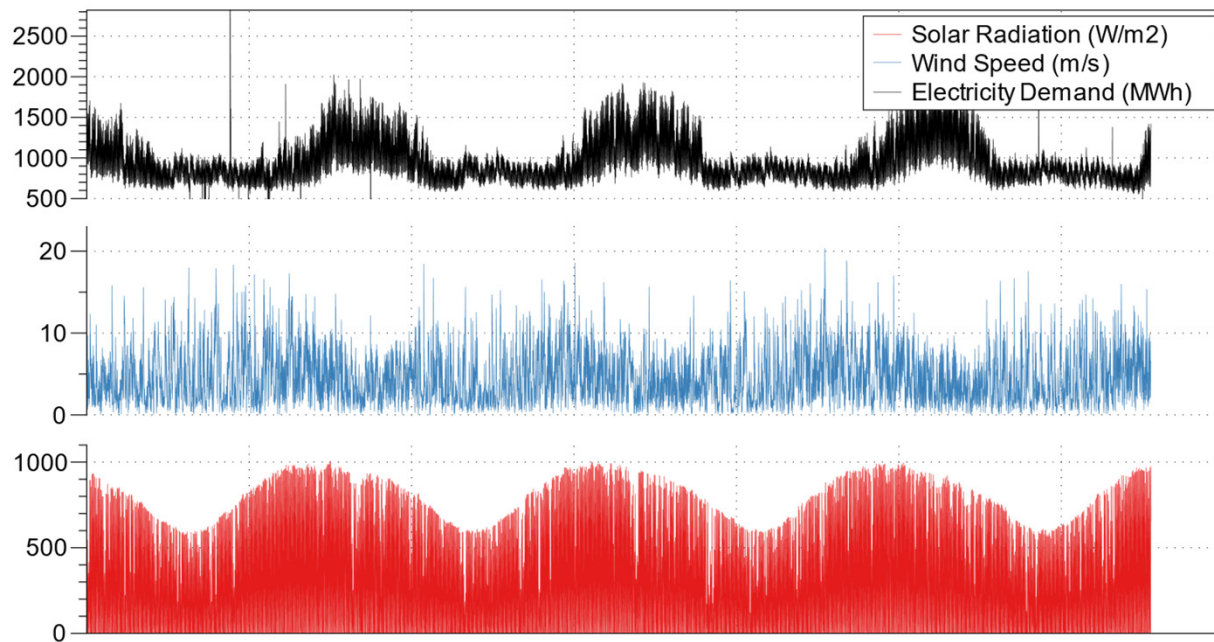
DIAGNOSIS OPTIMIZATION

EL PASO, TEXAS



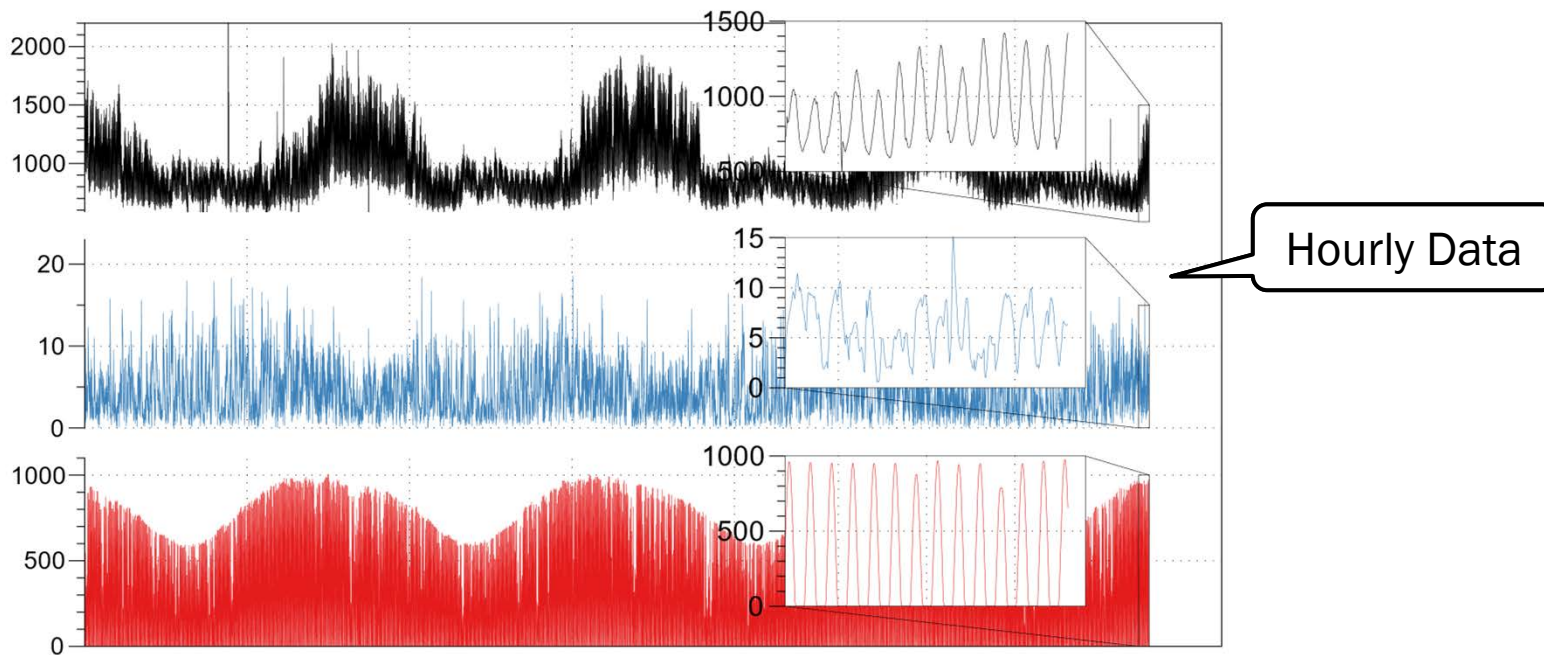
El Paso, Texas

Sun, Wind, and Electricity Demand Data (August 2016 – May 2020)



El Paso, Texas

Sun, Wind, and Electricity Demand Data (August 2016 – May 2020)



Joint Probability Distribution



Month



Hour



Temperature



Wind Speed



Shortwave Radiation
(W/m²)

- Edit
- Rename
- Copy
- Delete
- Exclude
- Set as Target Node
- Monitor
- Charts
- Imputation
- Probabilities >
- Select >
- Connect >
- Alignment >
- Properties >

Charts

Chart Settings

- Bar Chart
- Occurrence Matrix
- Distribution Function
- Density Function
- Box Plot
- Scatter of Points (2D)**
- Colored Line Plot
- Scatter of Points (2+1D)
- Bubble Chart (2+1D)
- Bubble Chart (2+2D)
- Time Series
- Scatter of Points (3D)
- Scatter of Points (3+1D)
- Bubble Chart (3+1D)
- Bubble Chart (3+2D)

Parameter

Variable on x: Temperature

Variable on y: Energy Demand (MWh)

Color:

Point Jittering:

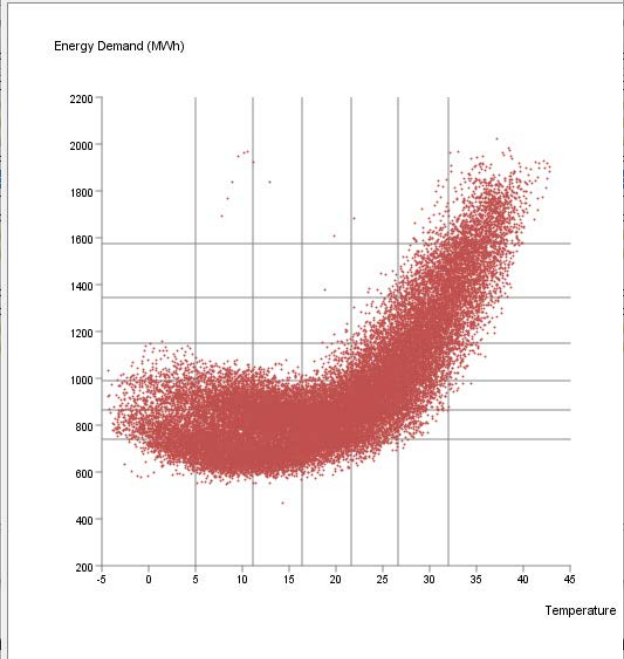
Point Size:

Close Show



Displayed Points: 32758
x: 41.75
y: 1735

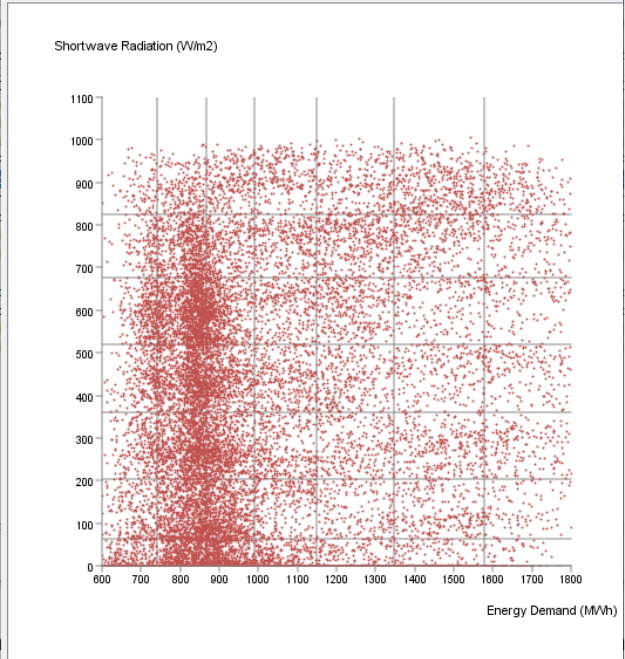
Energy Demand (MWh)



Close Save

Displayed Points: 32453
x: 1791
y: 209

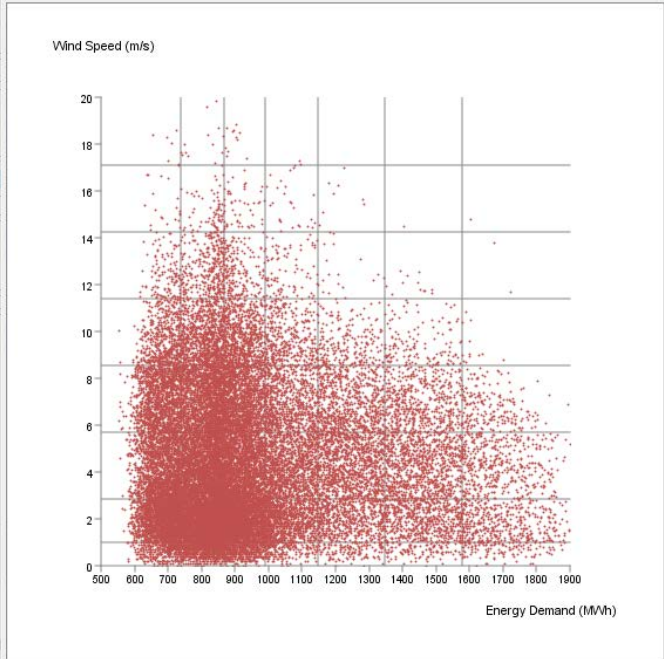
Shortwave Radiation (Win2)



Close Save

Displayed Points: 32721
x: 1896.5
y: 12

Wind Speed (m/s)



Close Save



Associated graph 6.xbl *



Month



Hour



Temperature



Shortwave Radiation
(W/m²)



Wind Speed (m/s)



Demand (MWh)

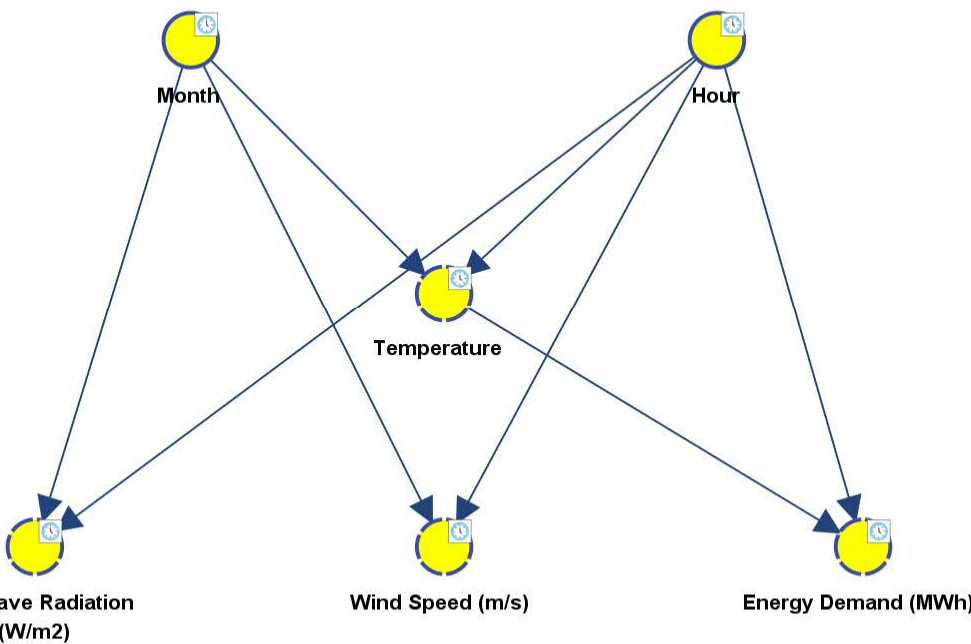
- Copy Selection
- Paste
- Paste Format
- Delete Selection
- Delete All Arcs
- Delete all Unfixed Arcs
- Delete all Unconnected Nodes
- Exclude all Unconnected Nodes
- Edit Structural Coefficient
- Edit Costs
- Edit Classes
- Edit Constants
- Edit Forbidden Arcs
- Edit Structural Priors
- Edit Temporal Indices**
- Edit Virtual Numbers of States
- Edit Reference States
- Edit Filtered States
- Edit Local Structural Coefficients
- Edit Number of Uniform Prior Samples
- Edit Experts
- Parameter Estimation with Trees
- Use Time Variable
- Properties >
- Save as Image...

Edit Temporal Indices ✕

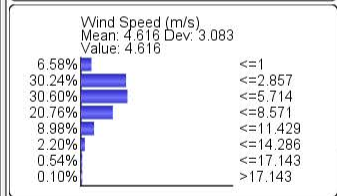
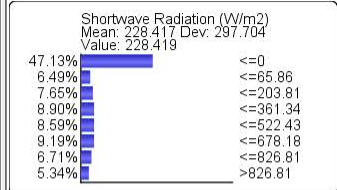
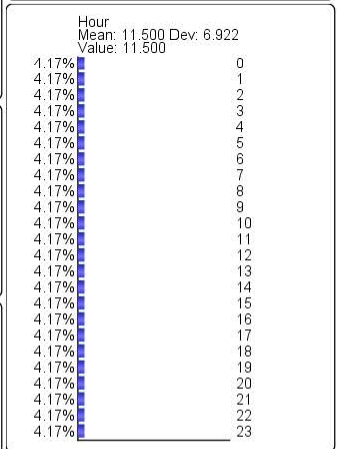
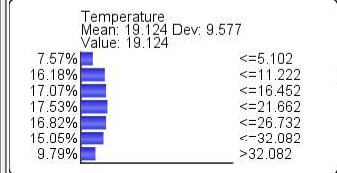
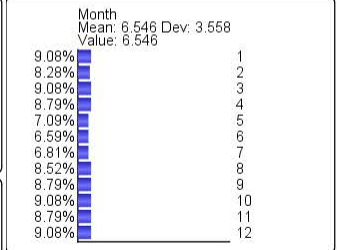
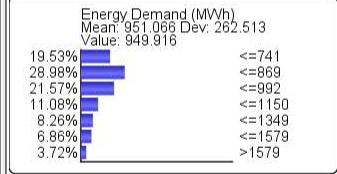
Nodes	Temporal Indices	
Energy Demand (MWh)	2	Associate Colors
Hour	0	Import
Month	0	Export
Shortwave Radiation (W/m ²)	1	
Temperature	1	
Wind Speed (m/s)	1	

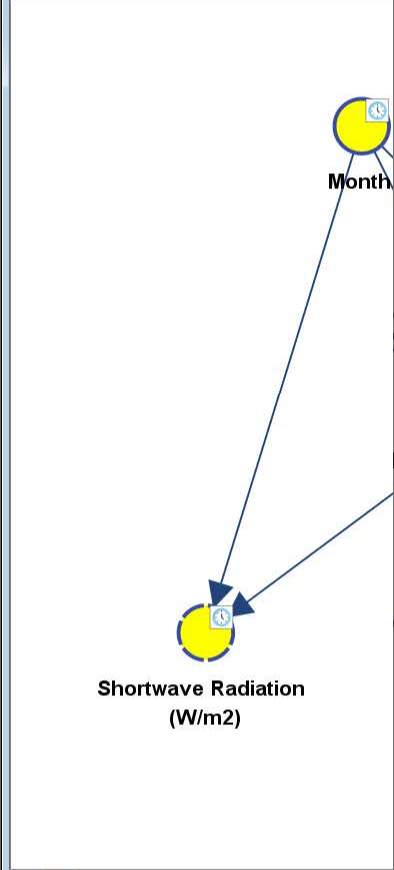


Associated graph 5.xbl

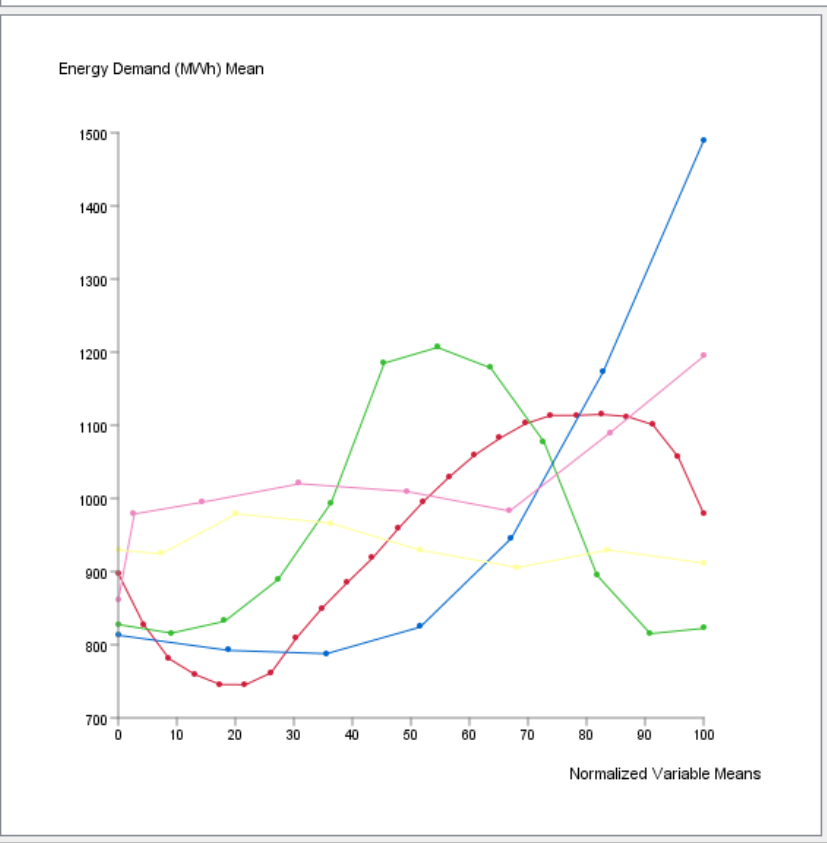


Joint Probability: 100.00%
 Log-Loss: 0
 Cases: 32,759
 Total Value: 1,220.121
 Mean Value: 203.353





Node:
 x:
 y:



Variables

- All Curves
- Hour
- Month
- Temperature
- Shortwave Radiation (W/m2)
- Wind Speed (m/s)

Month
 Mean: 6.546 Dev: 3.558
 Value: 6.546

9.08%	1
8.28%	2
9.08%	3
8.79%	4
7.09%	5
6.59%	6
6.81%	7
8.52%	8
8.79%	9
9.08%	10
8.79%	11
9.08%	12

Hour
 Mean: 11.500 Dev: 6.922
 Value: 11.500

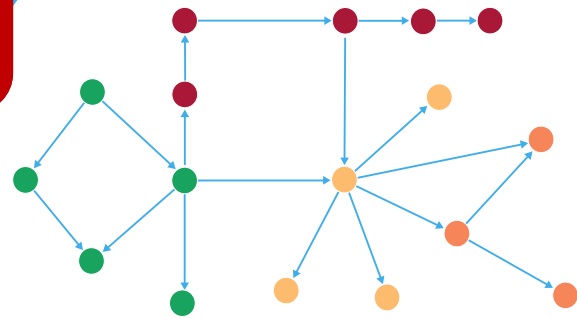
4.17%	0
4.17%	1
4.17%	2
4.17%	3
4.17%	4
4.17%	5
4.17%	6
4.17%	7
4.17%	8
4.17%	9
4.17%	10
4.17%	11
4.17%	12
4.17%	13
4.17%	14
4.17%	15
4.17%	16
4.17%	17
4.17%	18
4.17%	19
4.17%	20
4.17%	21
4.17%	22
4.17%	23

Reasoning Framework: Bayesian Networks

KNOWLEDGE MODELING

EXPERT
KNOWLEDGE

DATA
KNOWLEDGE DISCOVERY



**BAYESIAN
NETWORK**

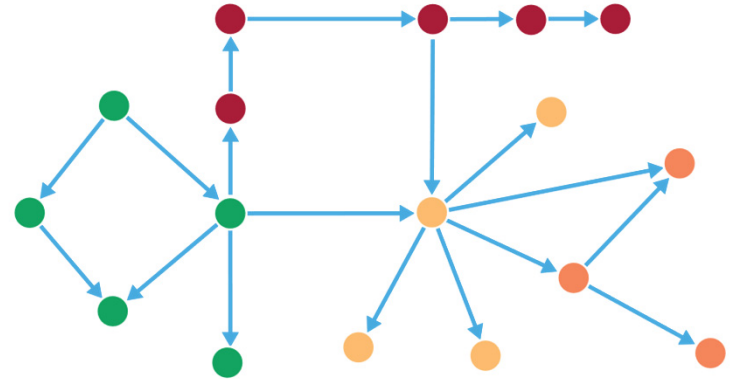
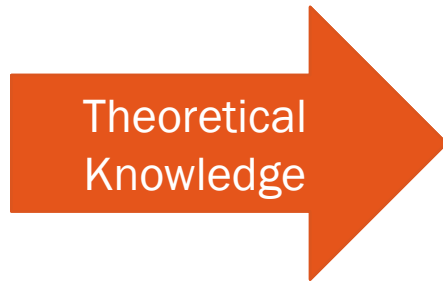
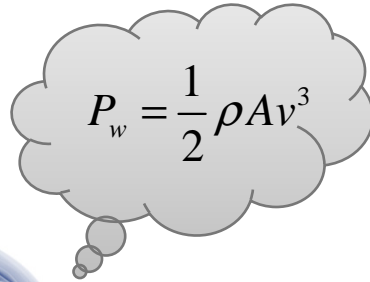
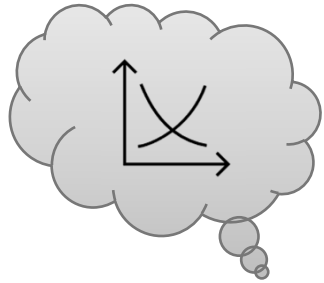
ANALYTICS SIMULATION

DECISION SUPPORT

RISK
MANAGEMENT

DIAGNOSIS OPTIMIZATION

Domain Knowledge



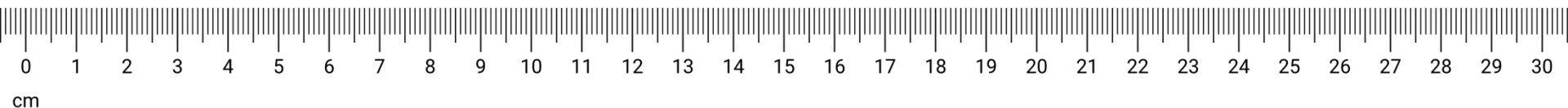
Domain Knowledge

Units

- Speed: m/s
- Power: MW
- Energy: MWh
- Power Density: W/m²
- Energy Density: MWh/t
- Temperature: ° Celsius

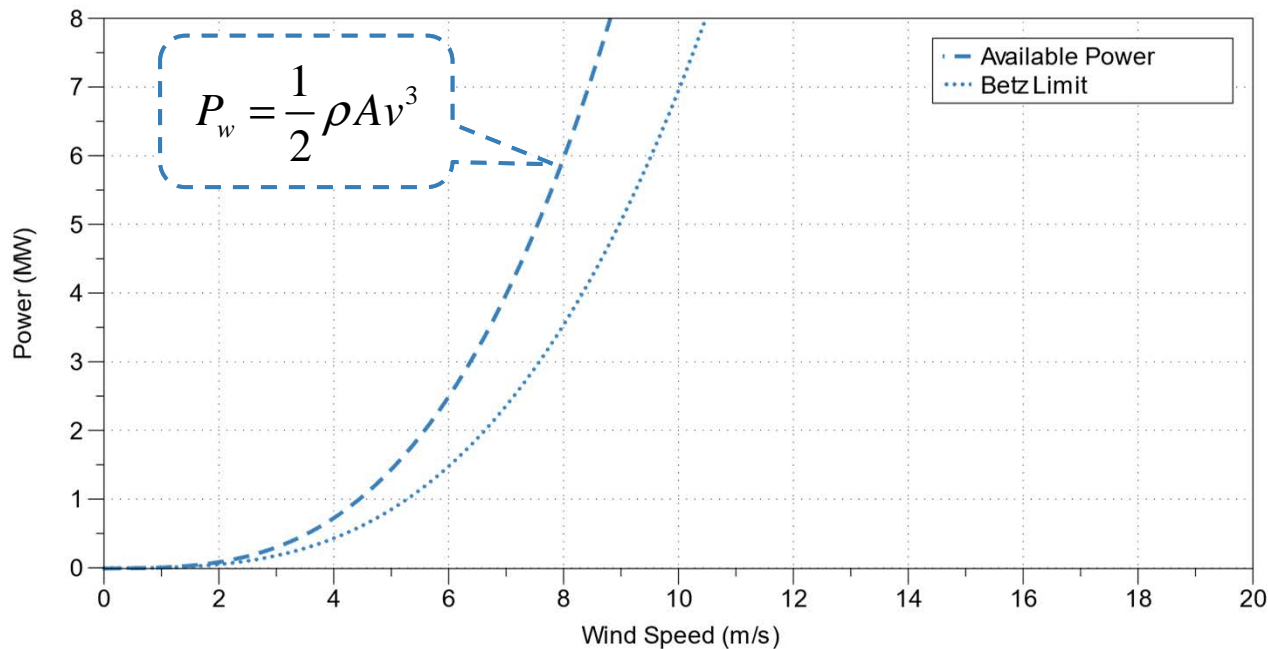


All Metric Units!



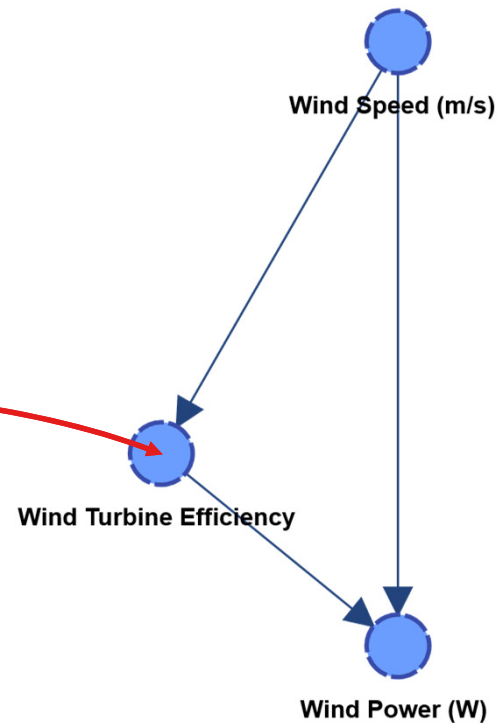
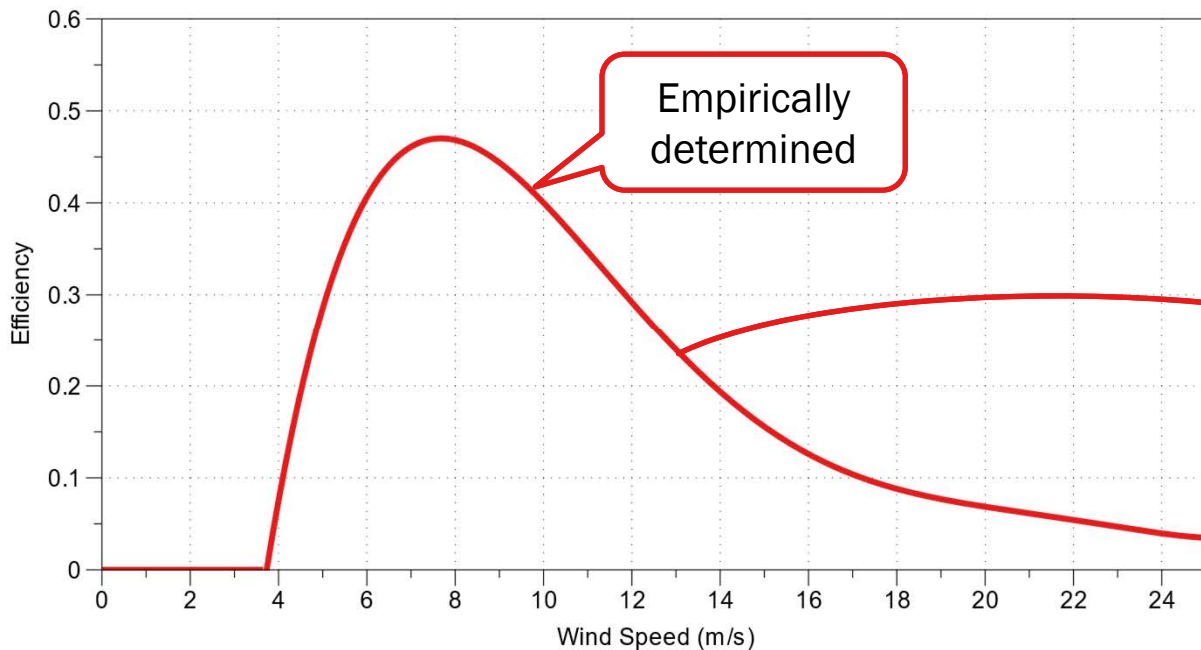
Domain Knowledge

Available Wind Energy



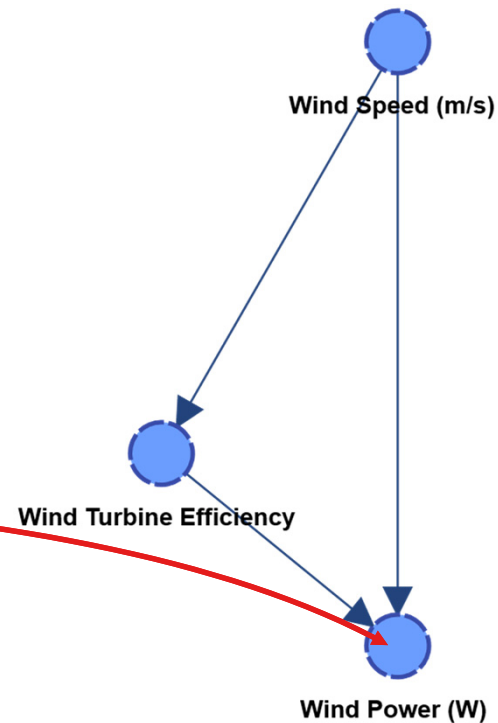
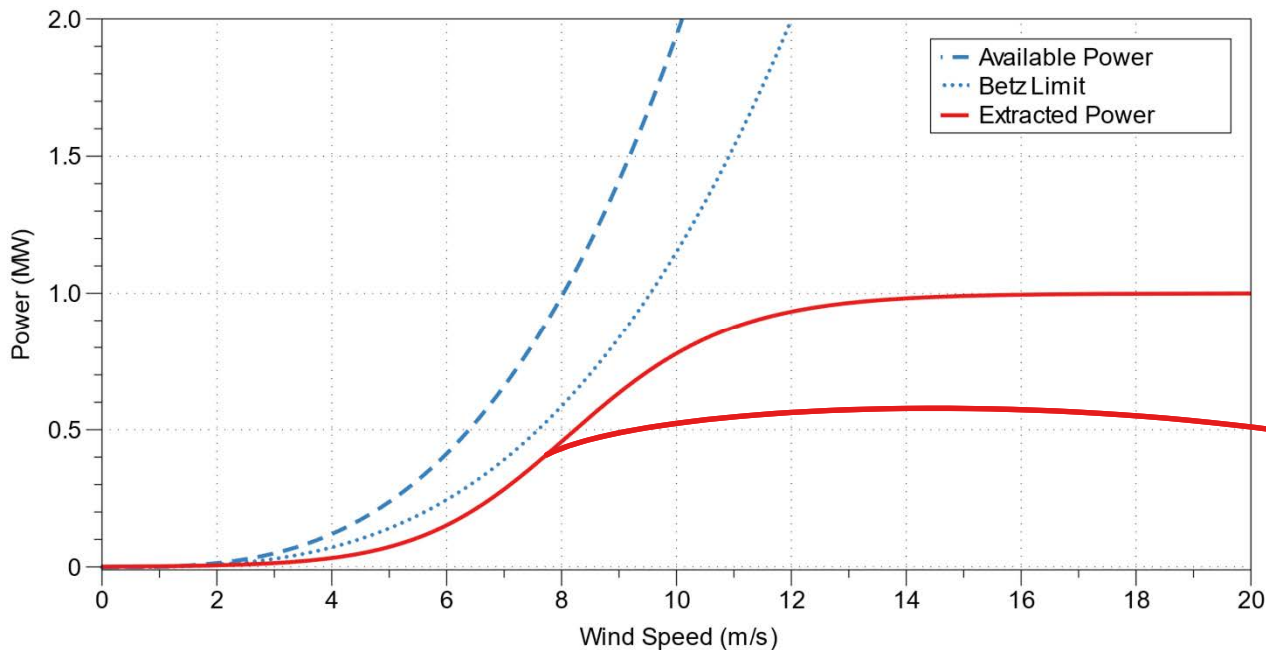
Domain Knowledge

Wind Turbine Efficiency



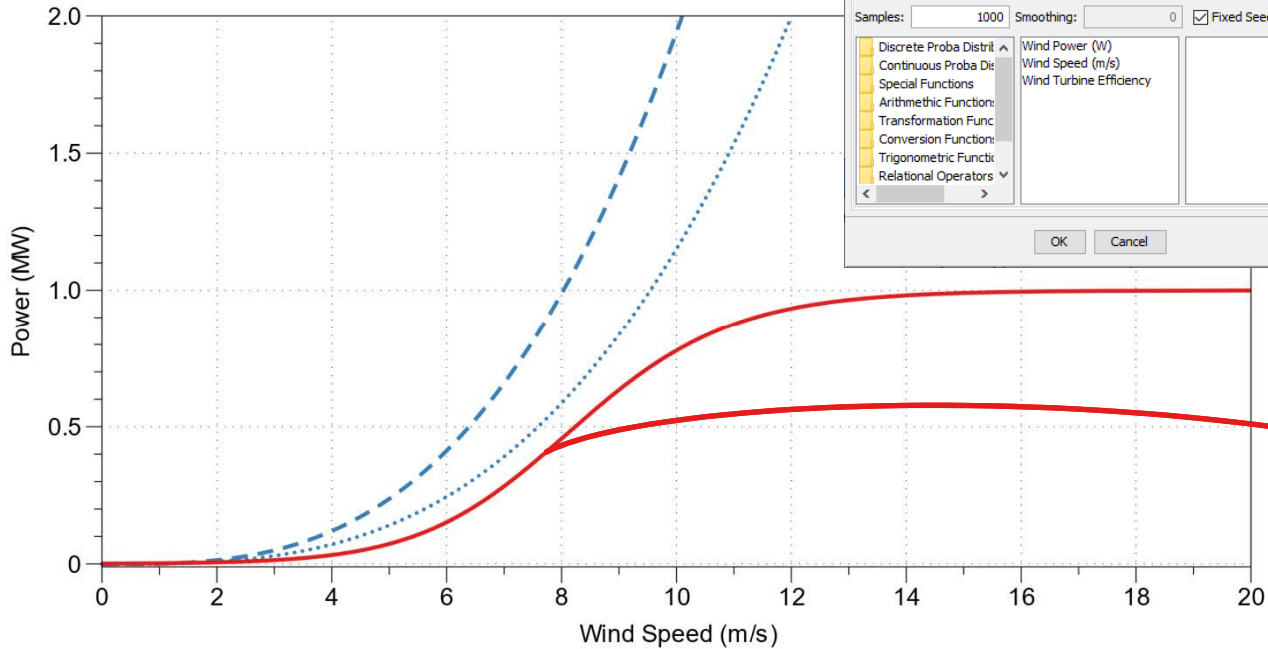
Domain Knowledge

Wind Energy



Domain Knowledge

Wind Energy



Node Editor

Node Selection: Wind Power (W) [Rename]

Reference State	Filtered State	Comment	Rendering Properties
States	Probability Distribution	Properties	Classes
States	Probability Distribution	Properties	Classes
States	Probability Distribution	Properties	Classes

Probabilistic Deterministic Tree Equation Updating

Equation Type: Deterministic Probabilistic

?Wind Power (W)? =
 $0.5 * 1.225 * \text{Wind Speed (m/s)}^3 * \text{Wind Turbine Efficiency?}$

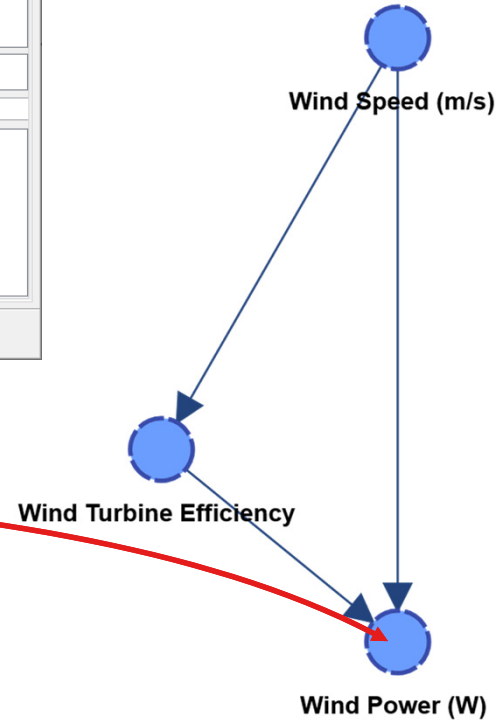
Please validate formula!

Samples: 1000 Smoothing: 0 Fixed Seed:

- Discrete Proba Distr
- Continuous Proba Di
- Special Functions
- Arithmetic Function
- Transformation Func
- Conversion Function
- Trigonometric Functi
- Relational Operators

Wind Power (W)
Wind Speed (m/s)
Wind Turbine Efficiency

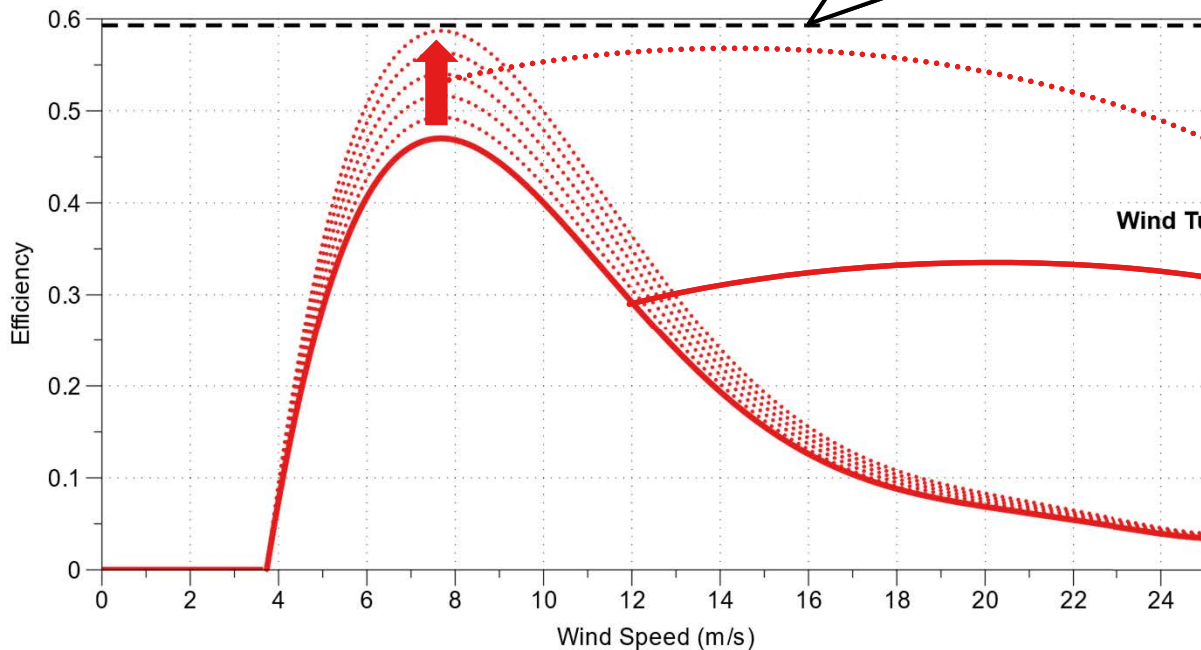
OK Cancel



Domain Knowledge

Future Wind Turbine Efficiency?

Betz Limit: 59.3%



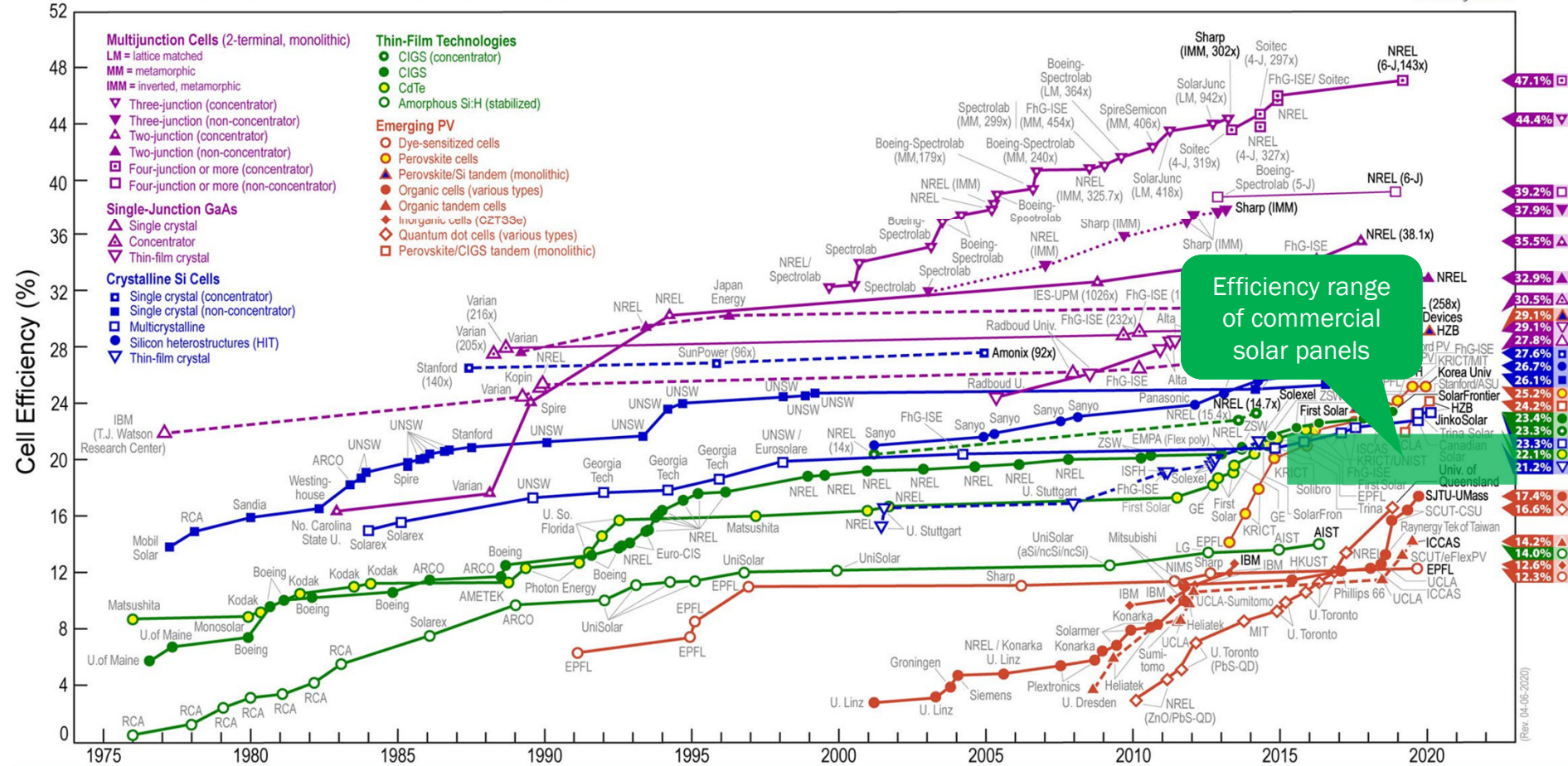
Wind Turbine Efficiency Factor

Wind Turbine Efficiency

Wind Speed (m/s)

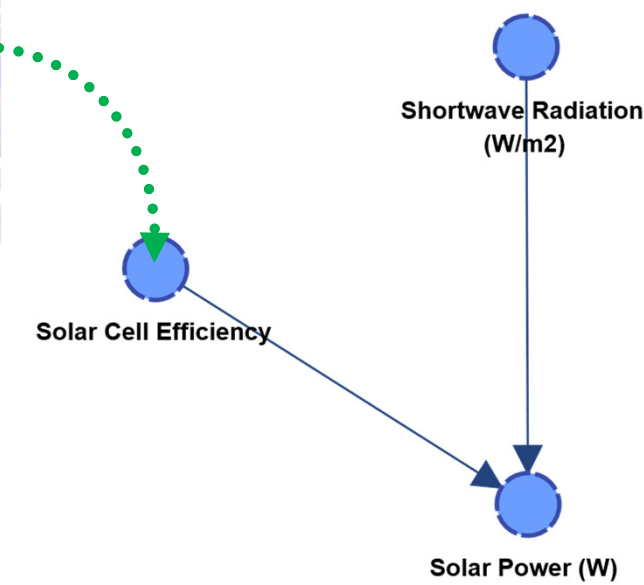
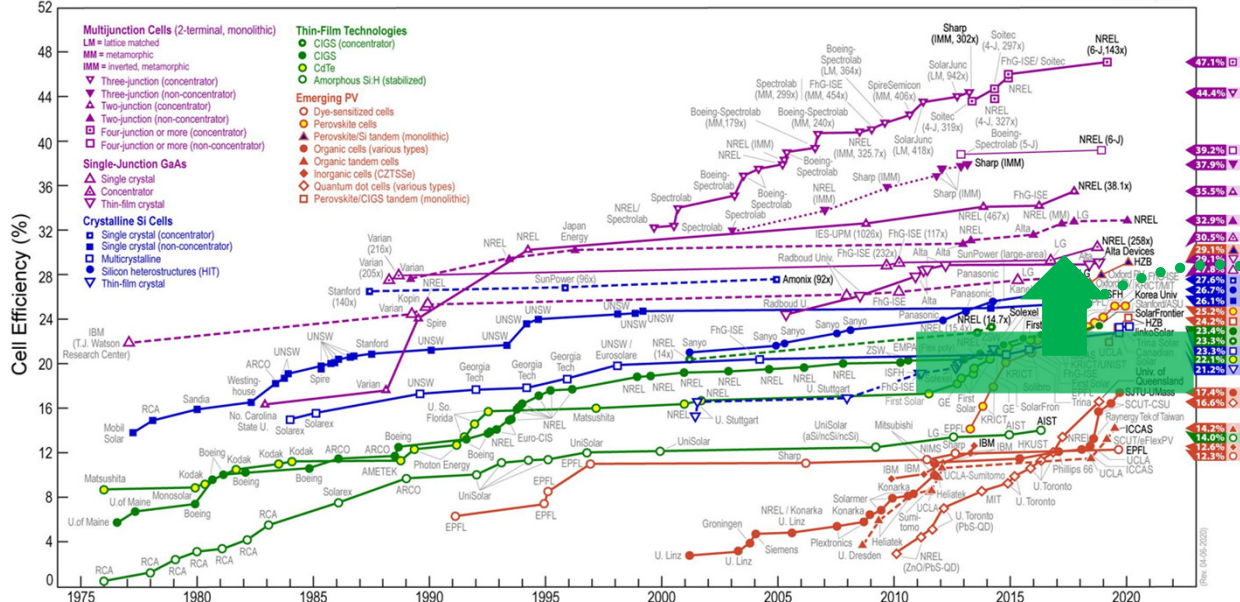
Wind Power (W)

Best Research-Cell Efficiencies



(Rev: 04-05-2020)

Best Research-Cell Efficiencies

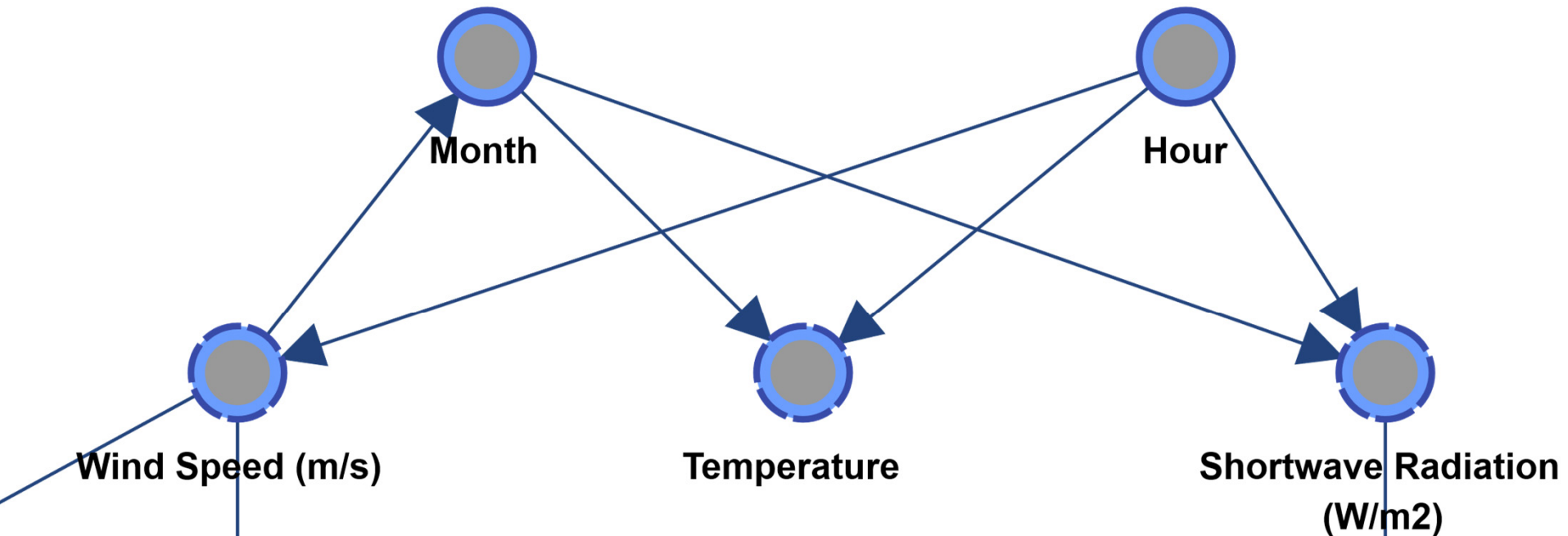


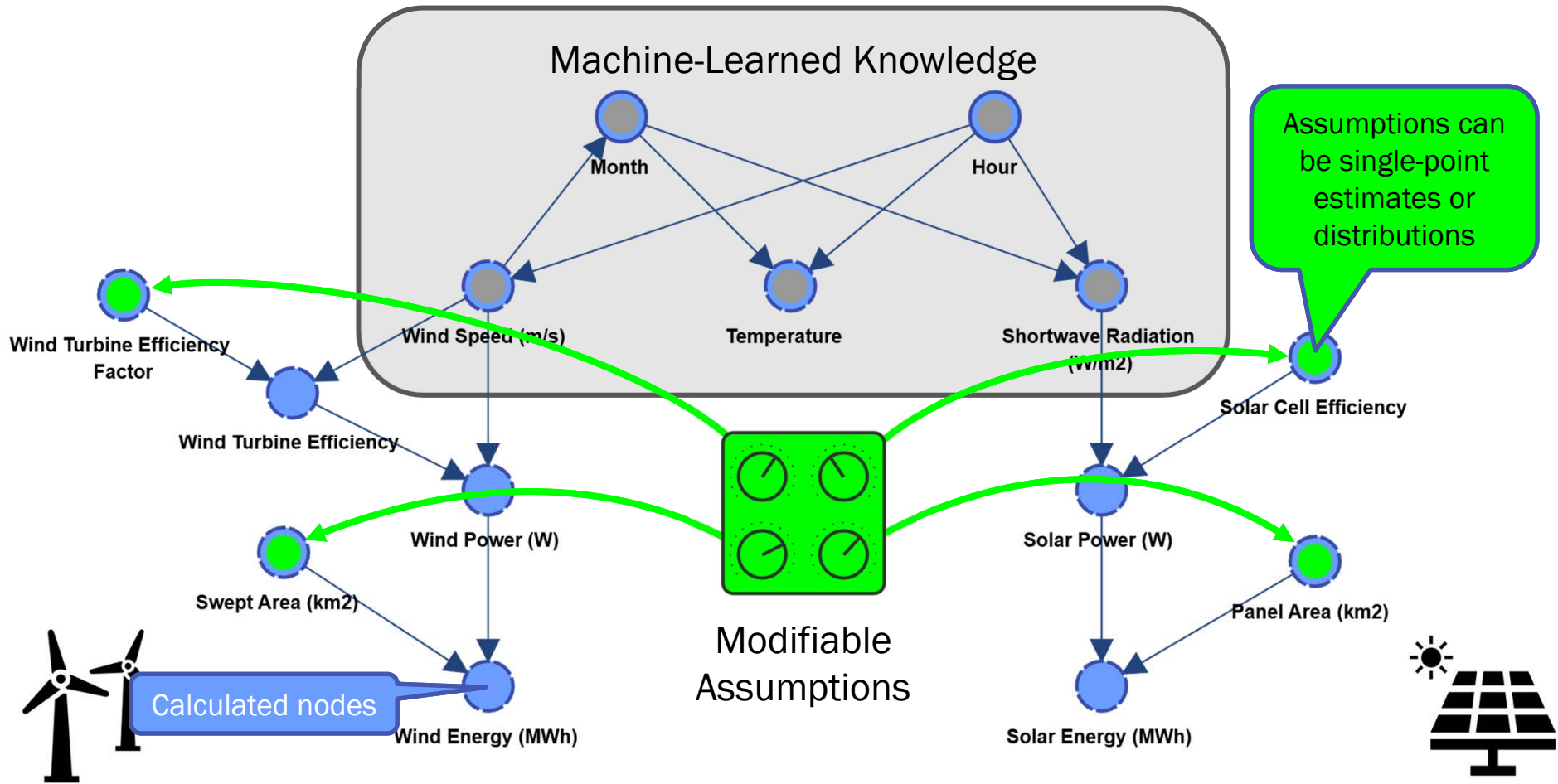
Domain Knowledge

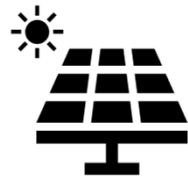
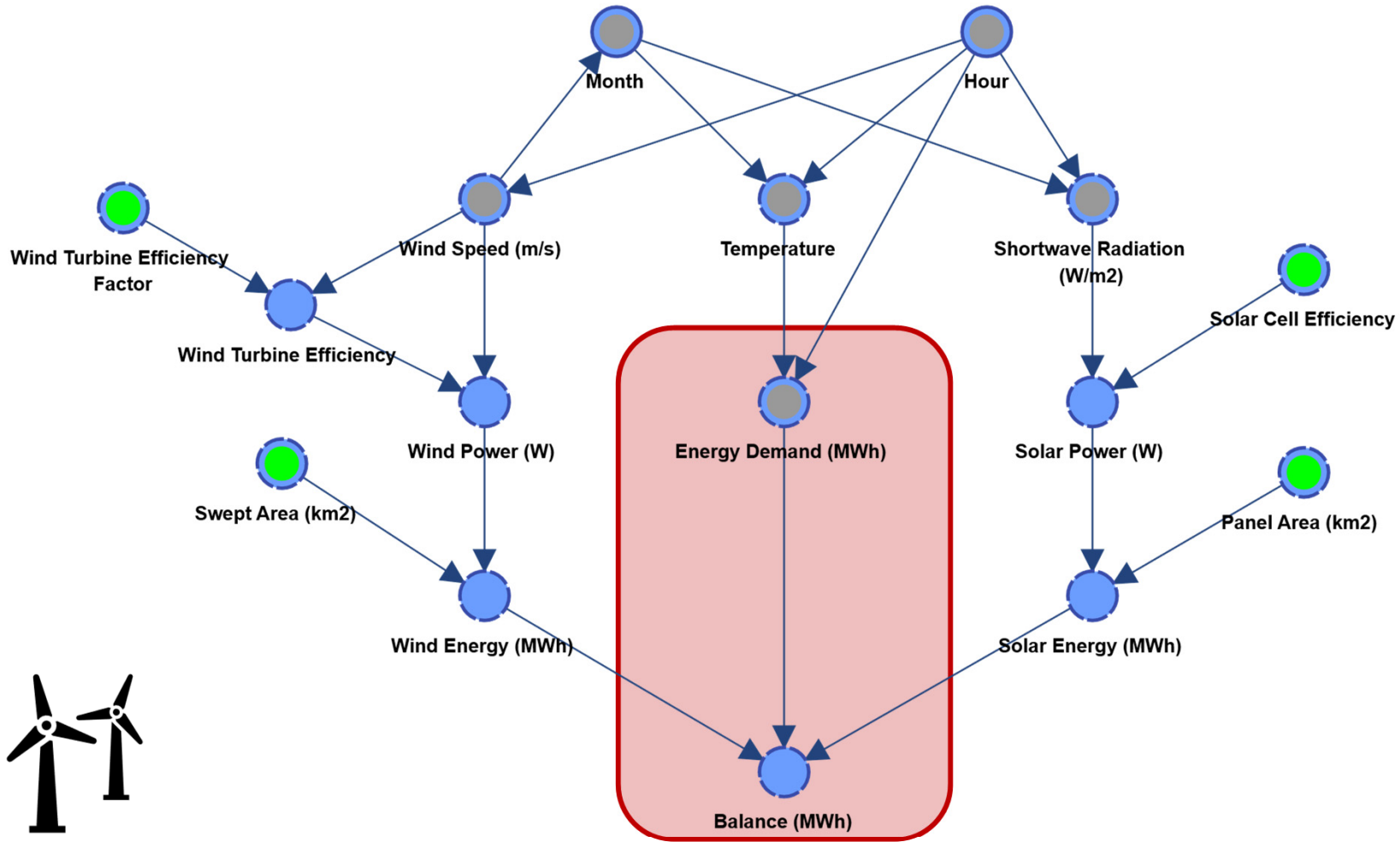
Modeling Assumptions

- Surface area of solar panels and swept area of wind turbines are infinitely scalable.
- Solar panel efficiency is constant.
- Wind turbine efficiency is wind speed-dependent.
- Gas turbine efficiency is load dependent.

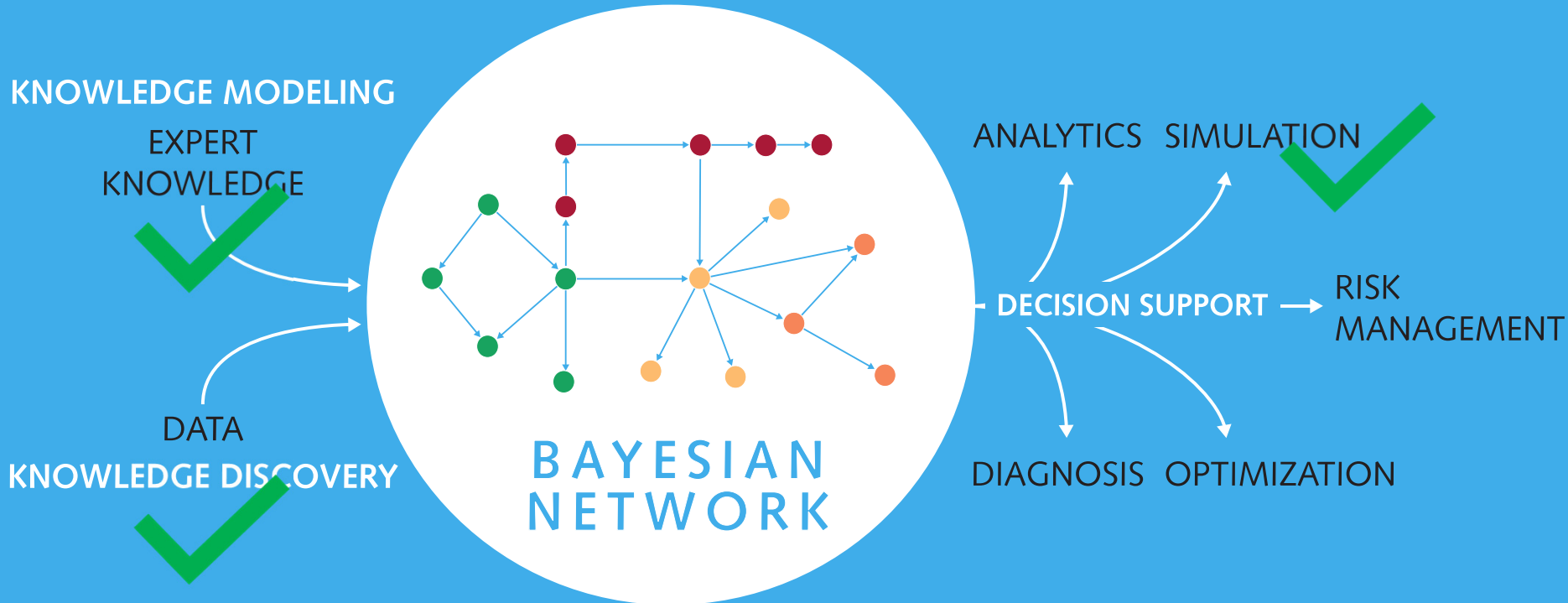
Machine-Learned Knowledge





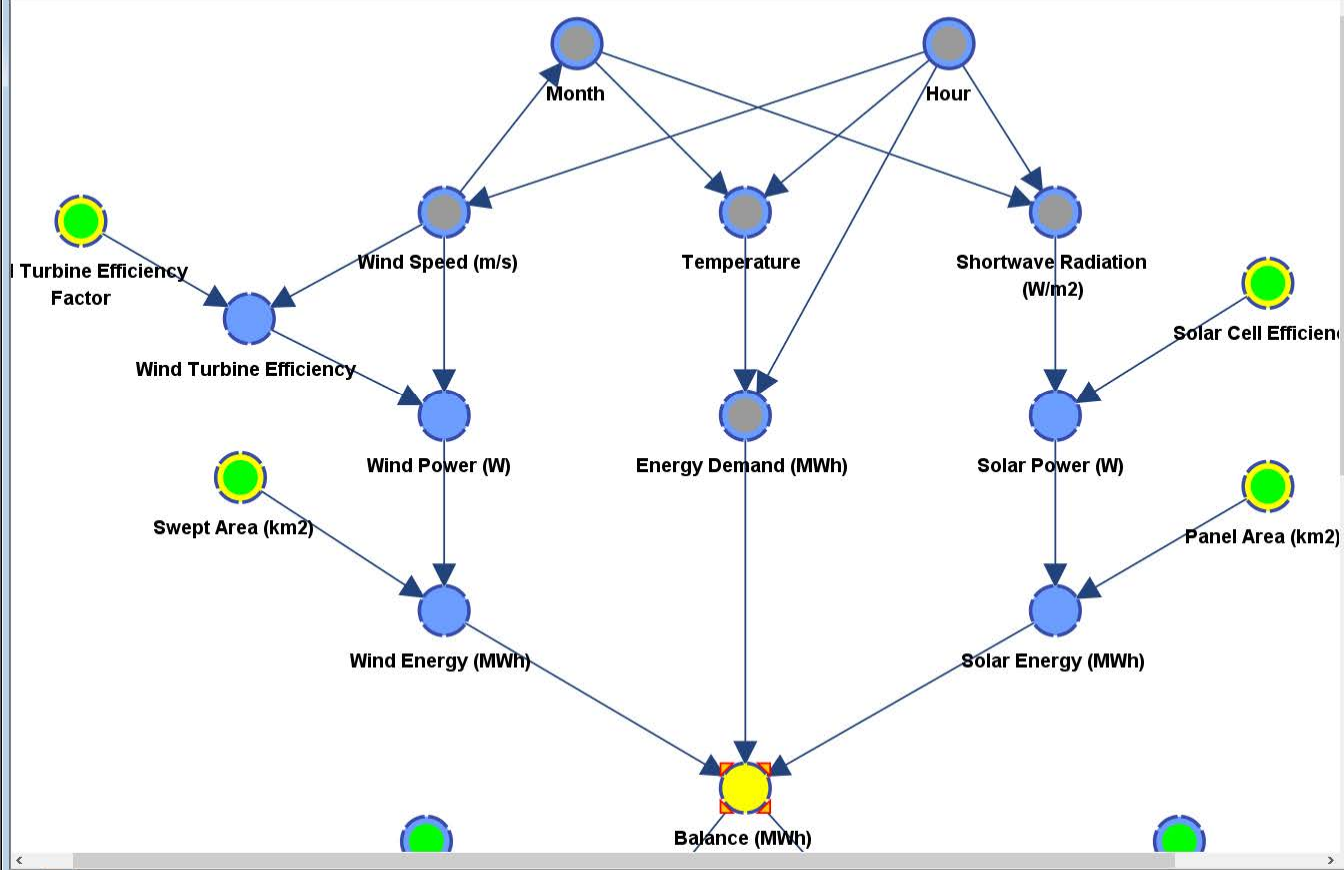


Reasoning Framework: Bayesian Networks





Energy Mix 7.xbl *



Joint Probability: 100.00%
 Log-Loss: 0
 Total Value: 90,166.840
 Mean Value: 3,606.674

Balance (MWh)
 Mean: -614.256 Dev: 698.311
 Value: -614.256

2.11%	<=-2250
0.33%	<=-2000
0.33%	<=-1750
0.77%	<=-1500
3.61%	<=-1250
8.43%	<=-1000
27.74%	<=-750
20.67%	<=-500
13.65%	<=-250
10.70%	<=0
5.13%	<=250
2.82%	<=500
1.60%	<=750
0.86%	<=1000
0.44%	<=1250
0.25%	<=1500
0.14%	<=1750
0.08%	<=2000
0.05%	<=2250
0.29%	>2250

Solar Cell Efficiency
 Mean: 0.325 Dev: 0.11
 Value: 0.325

14.29%	<=0
14.29%	<=0.25
14.29%	<=0.5
14.29%	<=0.75
14.29%	<=1
14.29%	<=1.25
14.29%	<=1.5
14.29%	<=1.75
14.29%	<=2
14.29%	>2

Panel Area (km2)
 Mean: 2.083 Dev: 1.59
 Value: 2.083

16.67%	<=0
16.67%	<=1
16.67%	<=2
16.67%	<=3
16.67%	<=4
16.67%	>4

Wind Turbine Efficiency Factor
 Mean: 1.104 Dev: 0.080
 Value: 1.104

16.67%	=1
16.67%	<=1.05
16.67%	<=1.1
16.67%	<=1.15
16.67%	<=1.2
16.67%	>1.2

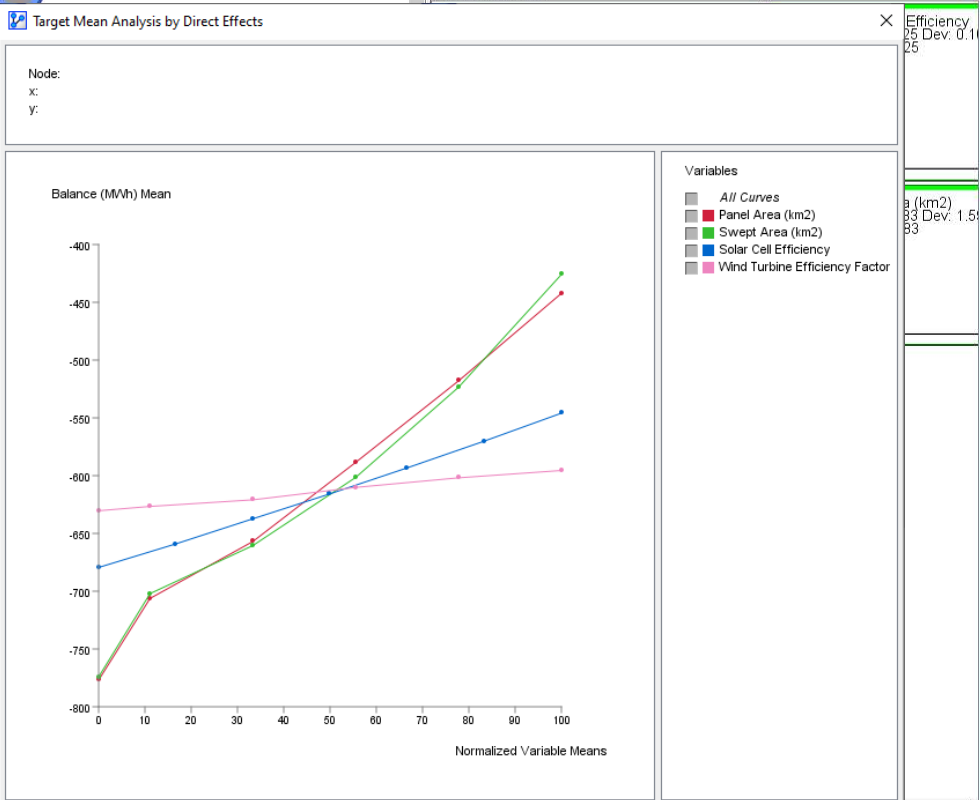
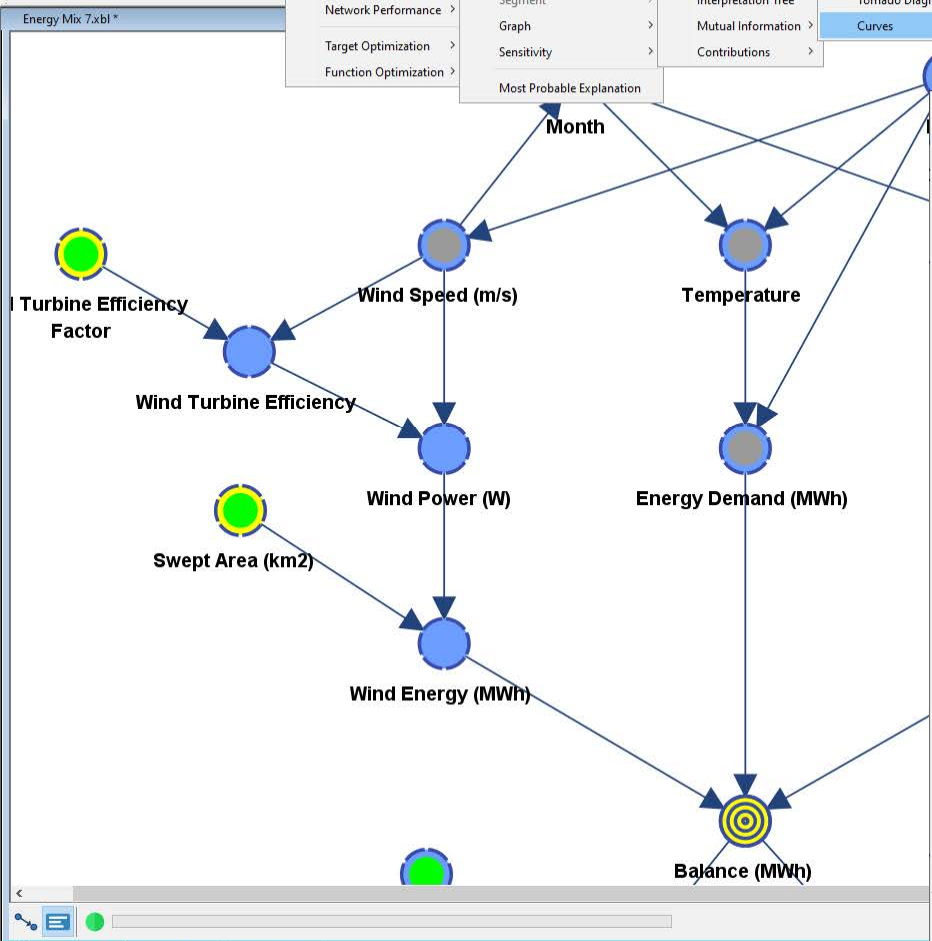
Swept Area (km2)
 Mean: 2.083 Dev: 1.592
 Value: 2.083

16.67%	<=0
16.67%	<=1
16.67%	<=2
16.67%	<=3
16.67%	<=4
16.67%	>4

- Visual >
- Report > Target > Target's Posterior >
- Network Performance > Segment > Interpretation Tree
- Target Optimization > Graph > Mutual Information >
- Function Optimization > Sensitivity > Contributions >
- Most Probable Explanation

- Histograms Shift+I
- Tornado Diagrams >
- Curves > Total Effects V
- Direct Effects

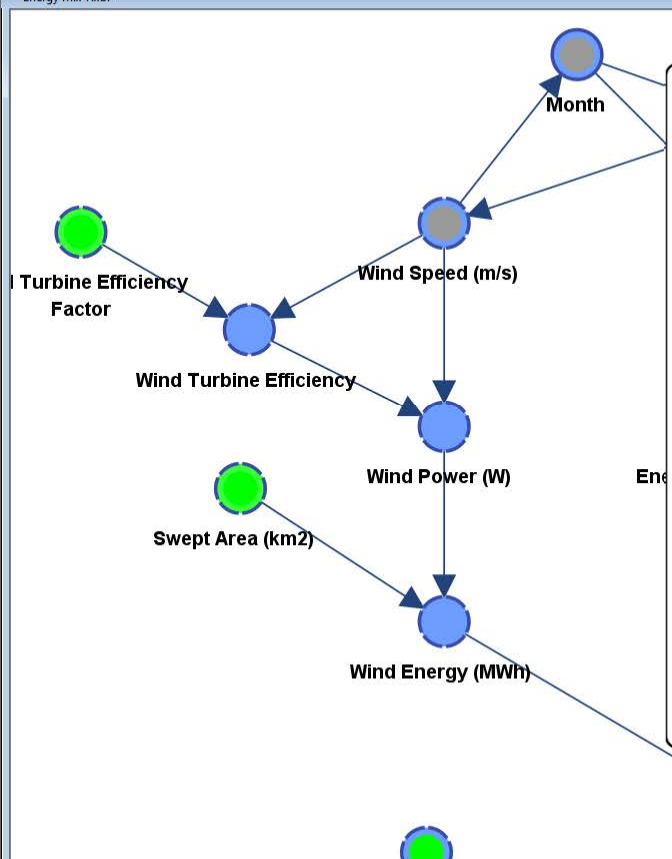
Joint Probability: 100.00%
Log-Loss: 0
Total Value: 90,166.840
Mean Value: 3,606.674



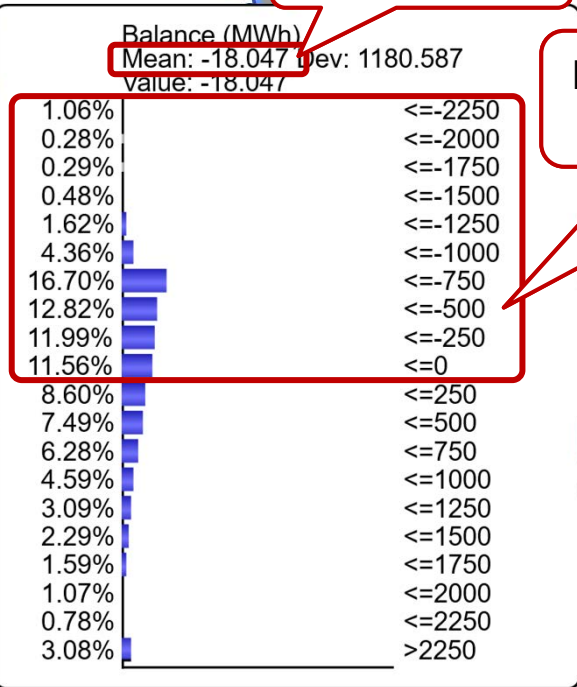
Close Save



Energy Mix 7.xbl *



Average ≈ 0



Negative balance in 61.2% of all time periods.

Joint Probability: 0.07%
 Log-Loss: 10.56
 Total Value: 68,212.247
 Mean Value: 2,728.490

Balance (MWh)

Solar Cell Efficiency
 Mean: 0.475 Dev: 0.0
 Value: 0.475

Panel Area (km2)
 Mean: 4.500 Dev: 0.0
 Value: 4.500

Wind Turbine Efficiency Factor
 Mean: 1.225 Dev: 0.000
 Value: 1.225

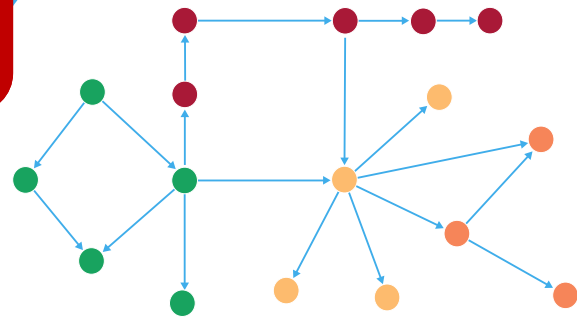
Swept Area (km2)
 Mean: 4.500 Dev: 0.000
 Value: 4.500

Reasoning Framework: Bayesian Networks

KNOWLEDGE MODELING

EXPERT
KNOWLEDGE

DATA
KNOWLEDGE DISCOVERY



**BAYESIAN
NETWORK**

ANALYTICS SIMULATION

DECISION SUPPORT

RISK
MANAGEMENT

DIAGNOSIS OPTIMIZATION

Battery Storage

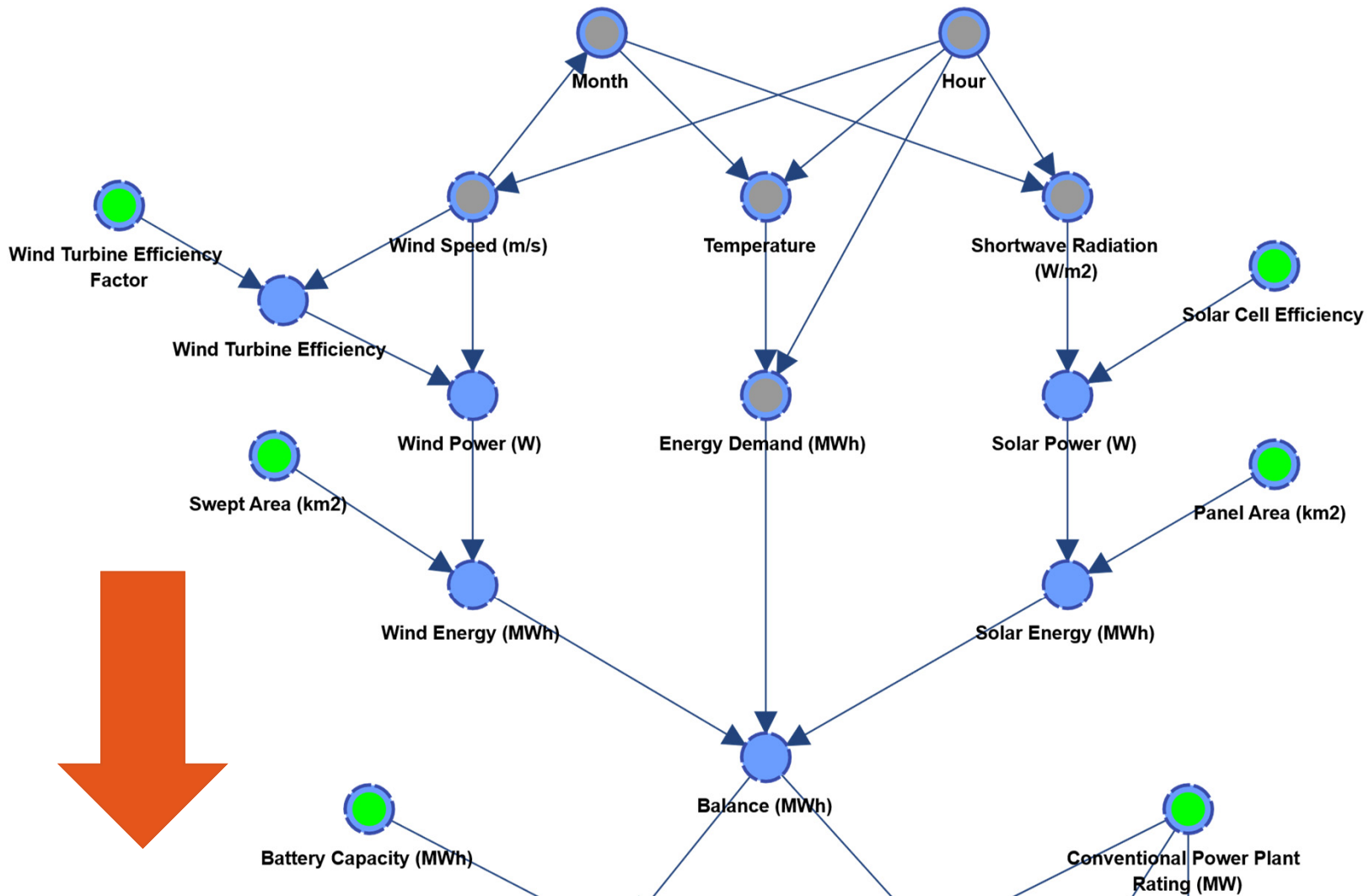


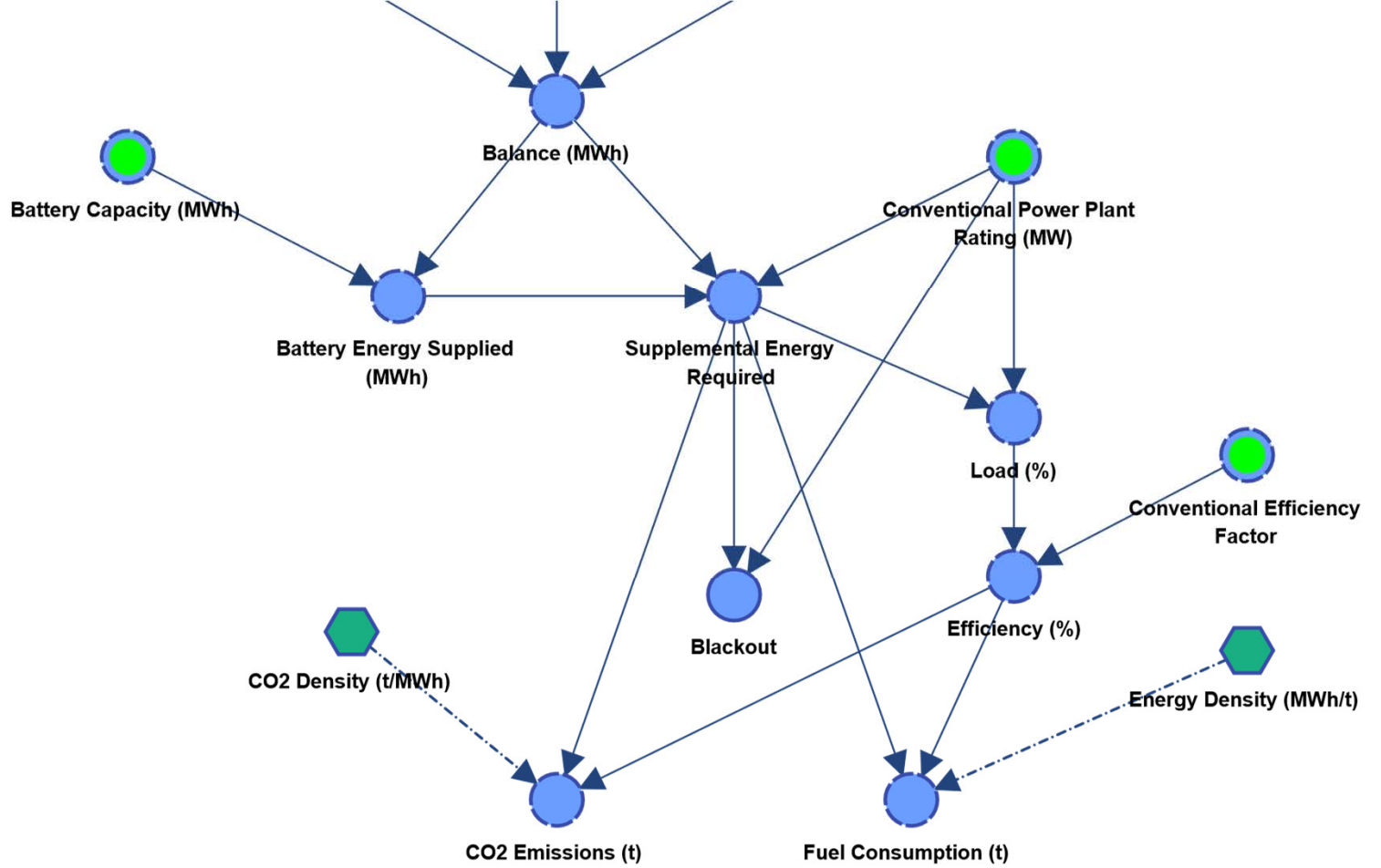
Caveat

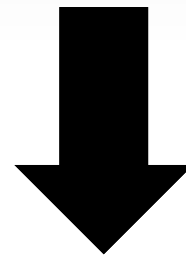
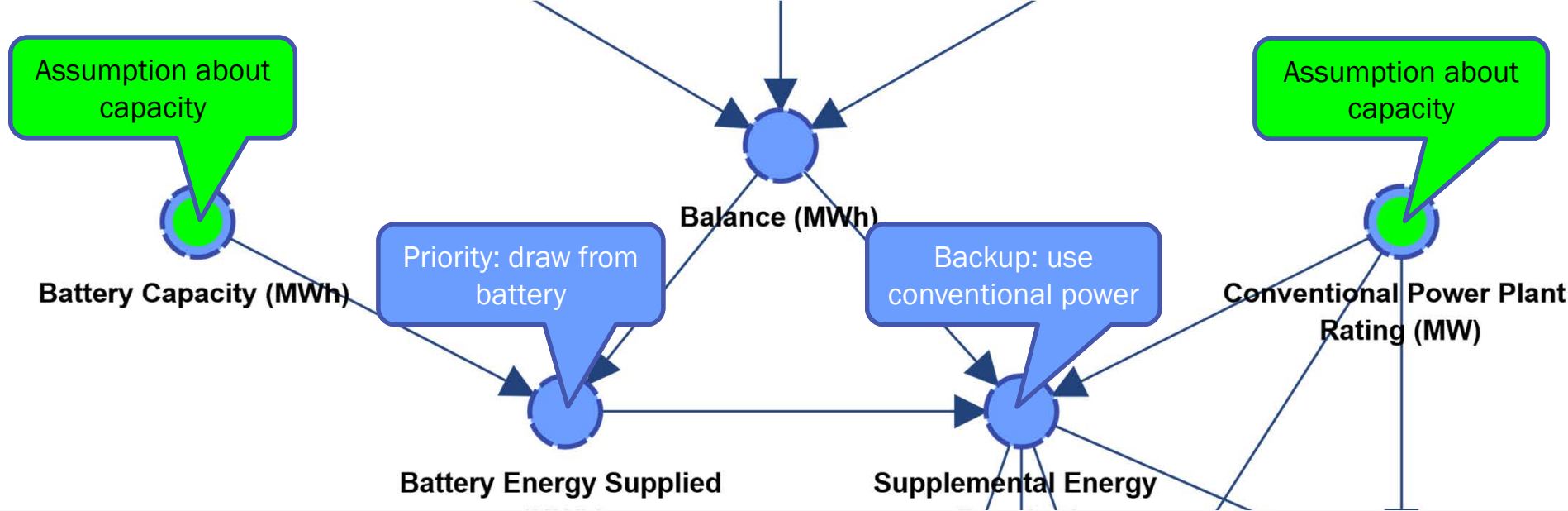


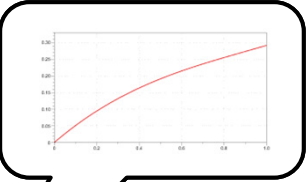
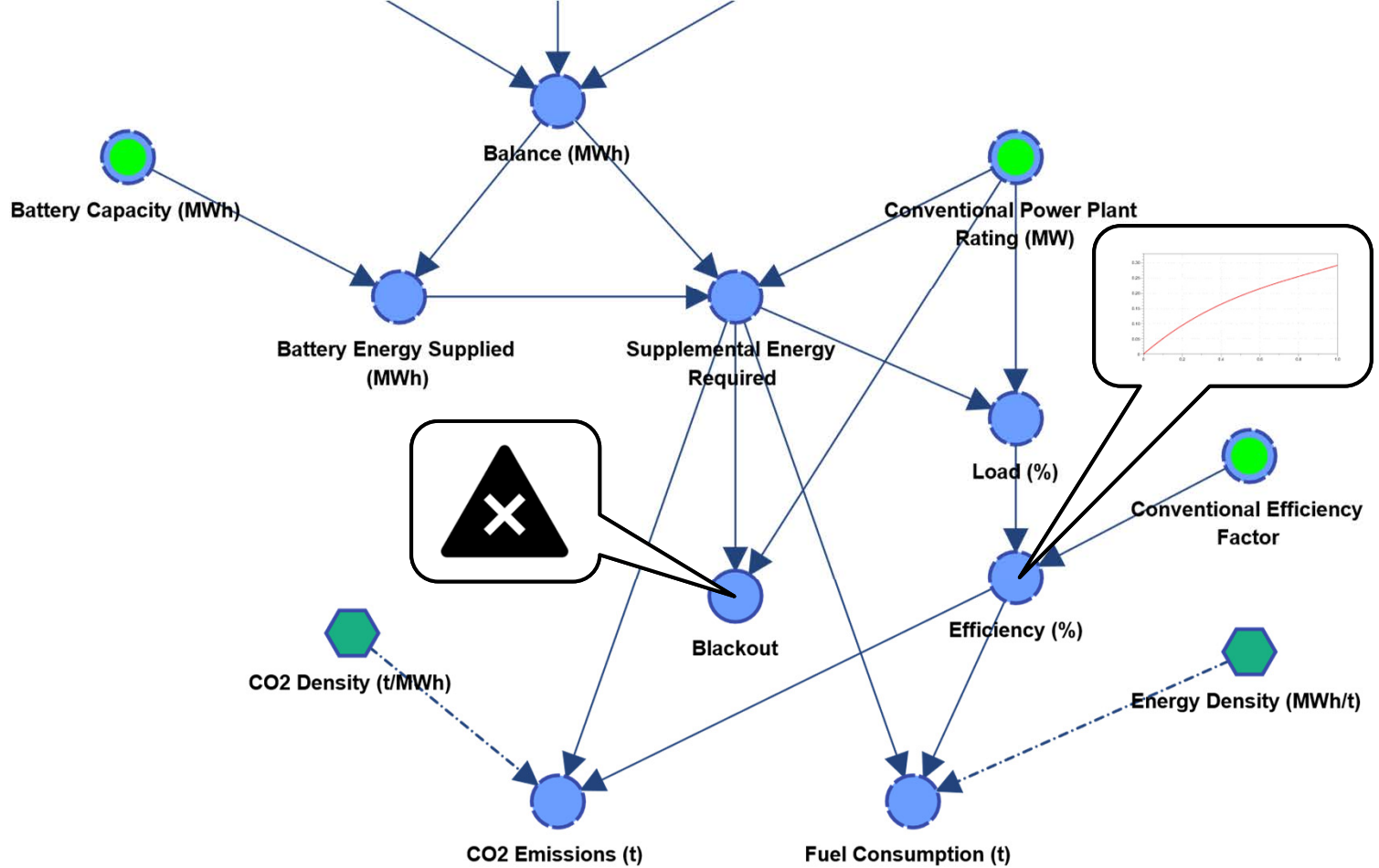
Conventional Power Plant

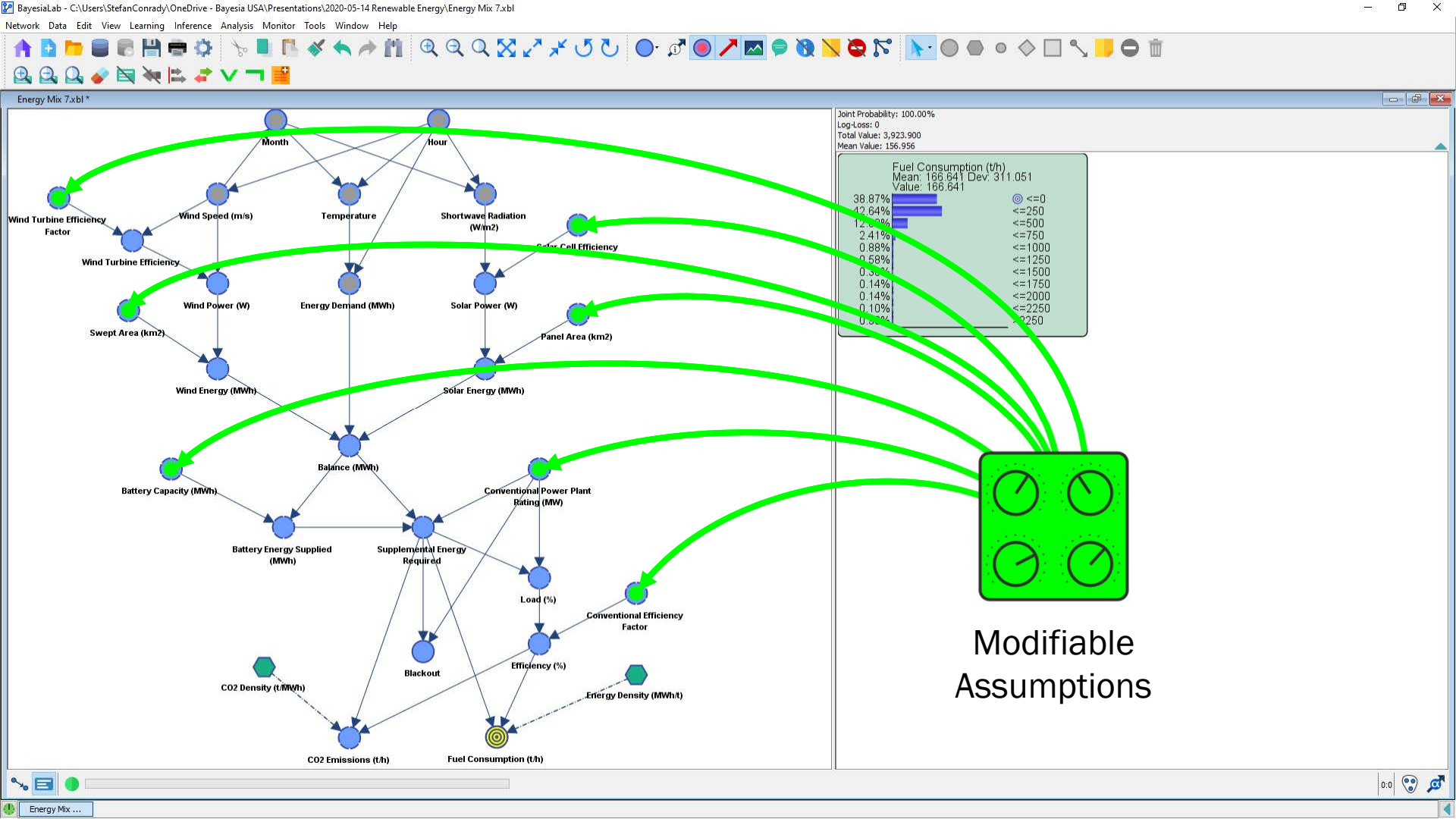




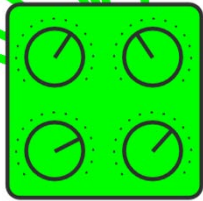
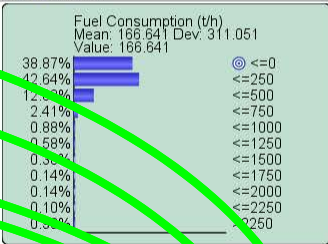








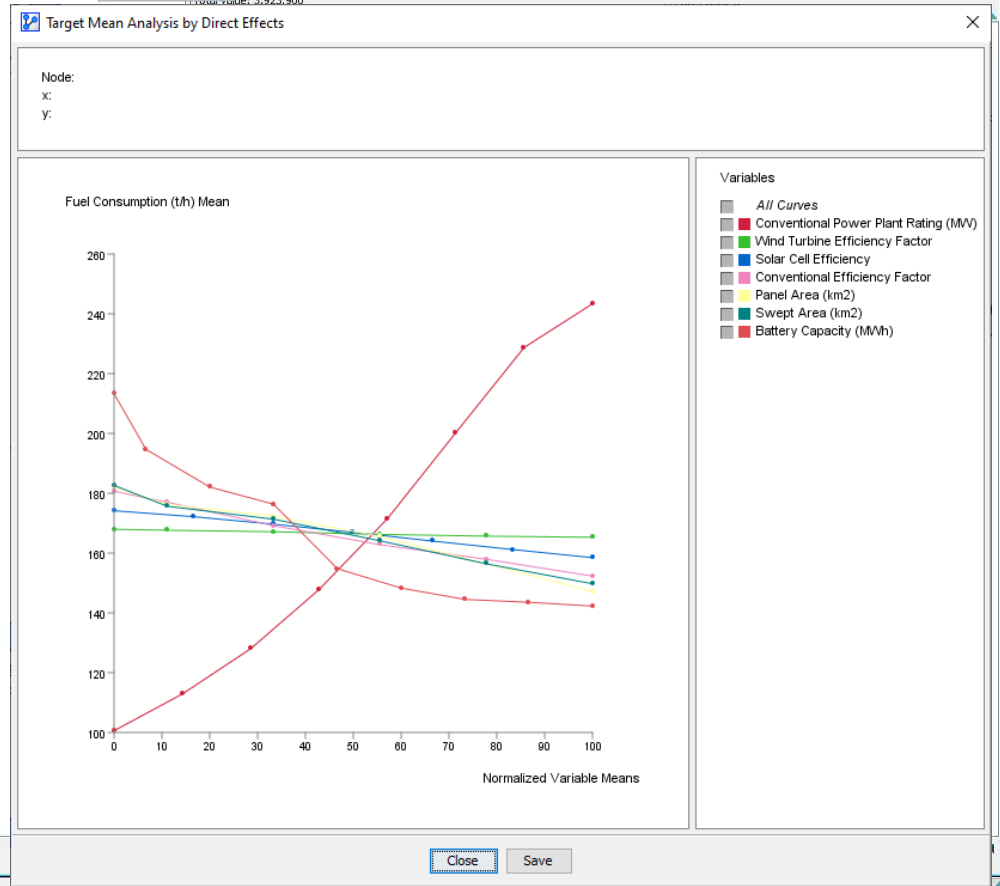
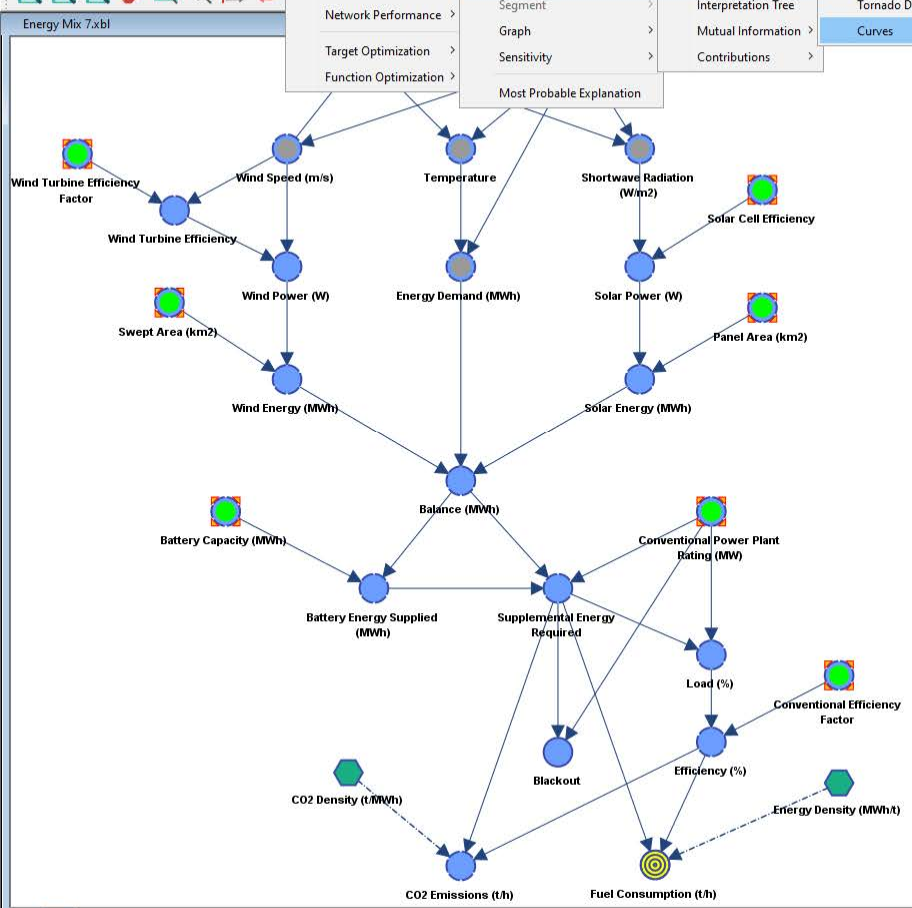
Joint Probability: 100.00%
 Log-Loss: 0
 Total Value: 3,923.900
 Mean Value: 156.956

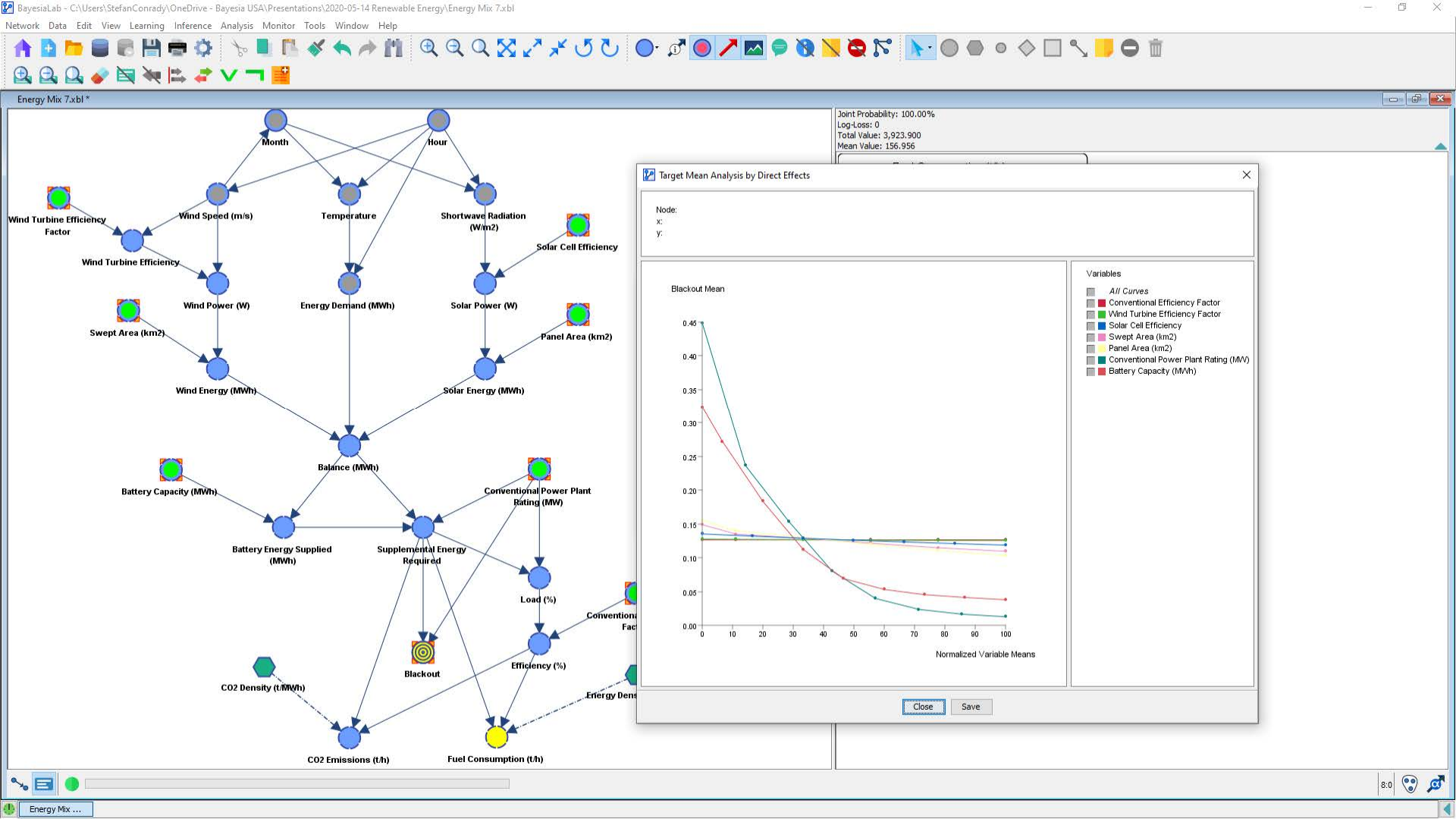


Modifiable
Assumptions

- Visual > Overall > Target > Target's Posterior > Histograms Shift+I
- Report > Target > Target's Posterior > Tornado Diagrams > Curves > Total Effects V
- Network Performance > Segment > Interpretation Tree
- Target Optimization > Graph > Mutual Information
- Function Optimization > Sensitivity > Contributions
- Most Probable Explanation

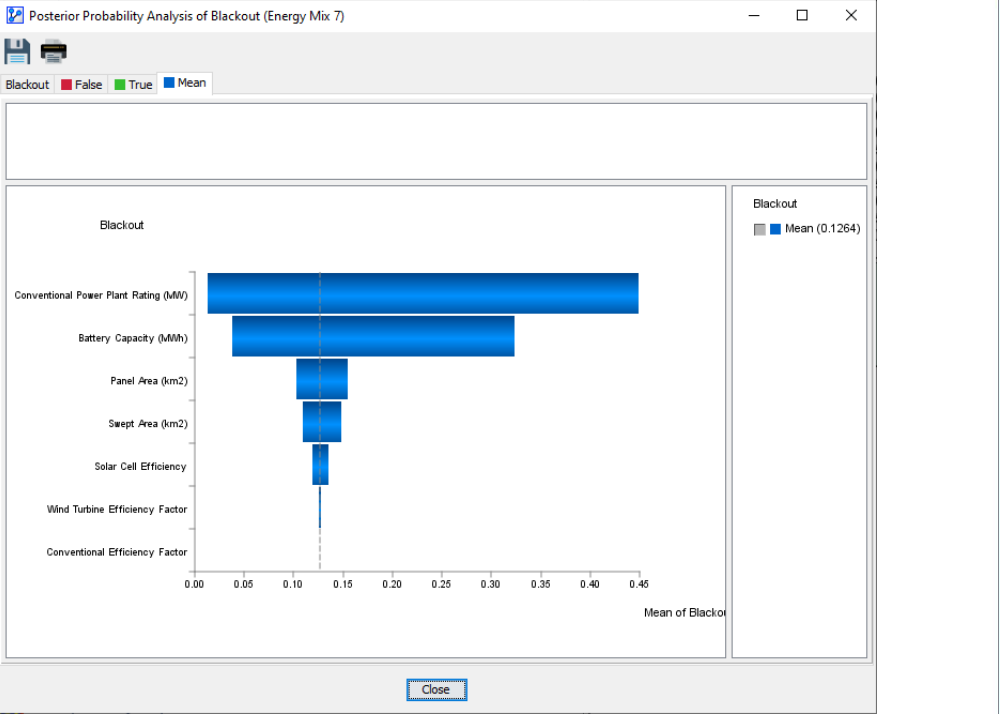
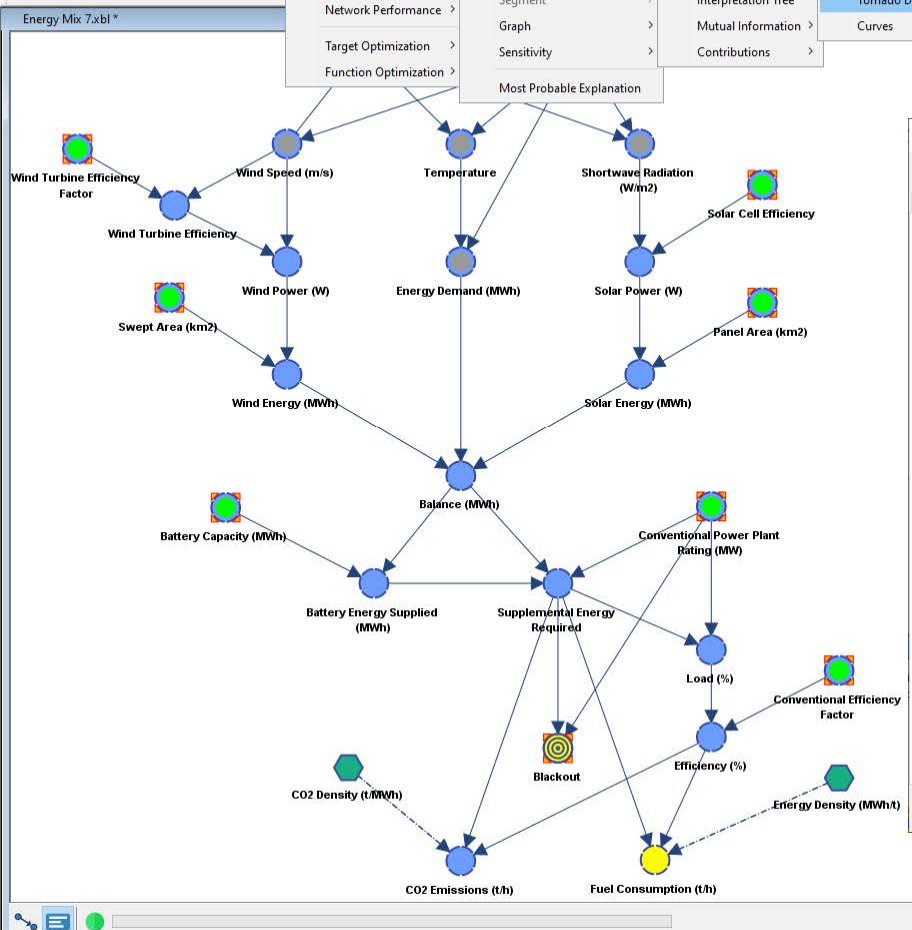
ability: 100.00%
 : 0
 : 3,923.900





- Visual > Overall > Target > Target's Posterior > Histograms Shift+I
- Report > Segment > Interpretation Tree
- Network Performance > Graph > Mutual Information > Tornado Diagrams > Total Effects
- Target Optimization > Sensitivity > Contributions > Curves > Direct Effects
- Function Optimization > Most Probable Explanation

Probability: 100.00%
 Log-Loss: 0
 Total Value: 3,923.900
 Mean Value: 156.956





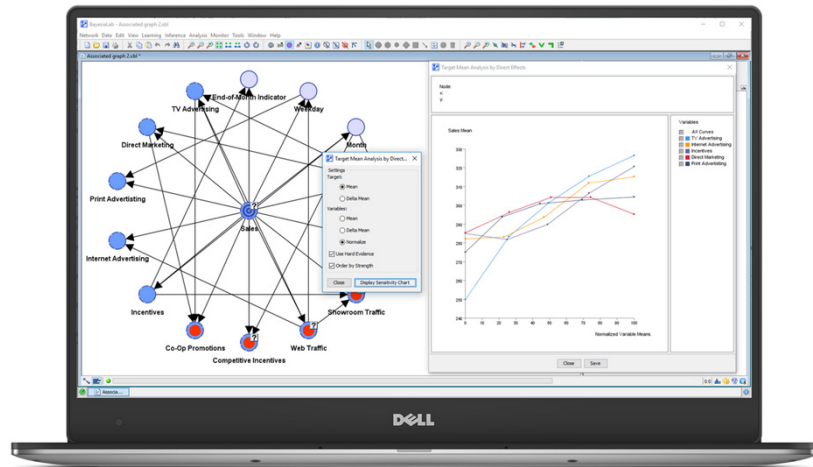
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In Conclusion...

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Reasoning Under Uncertainty

- March 20 Part 1 – Differential Diagnosis of Diseases
- March 26 Part 2 – Pandemic Triage with Bayesian Networks
- April 9 Part 3 – Epidemic Modeling with Temporal Bayesian Networks
- April 30 Part 4 – Representing Spatial and Temporal Dynamics
- May 14 Part 5 – Reasoning About Renewable Energy
- May 28 t.b.d.

Program Subject to Change – Check for Updates



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Thank you!



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