

Bayesian Network Queries

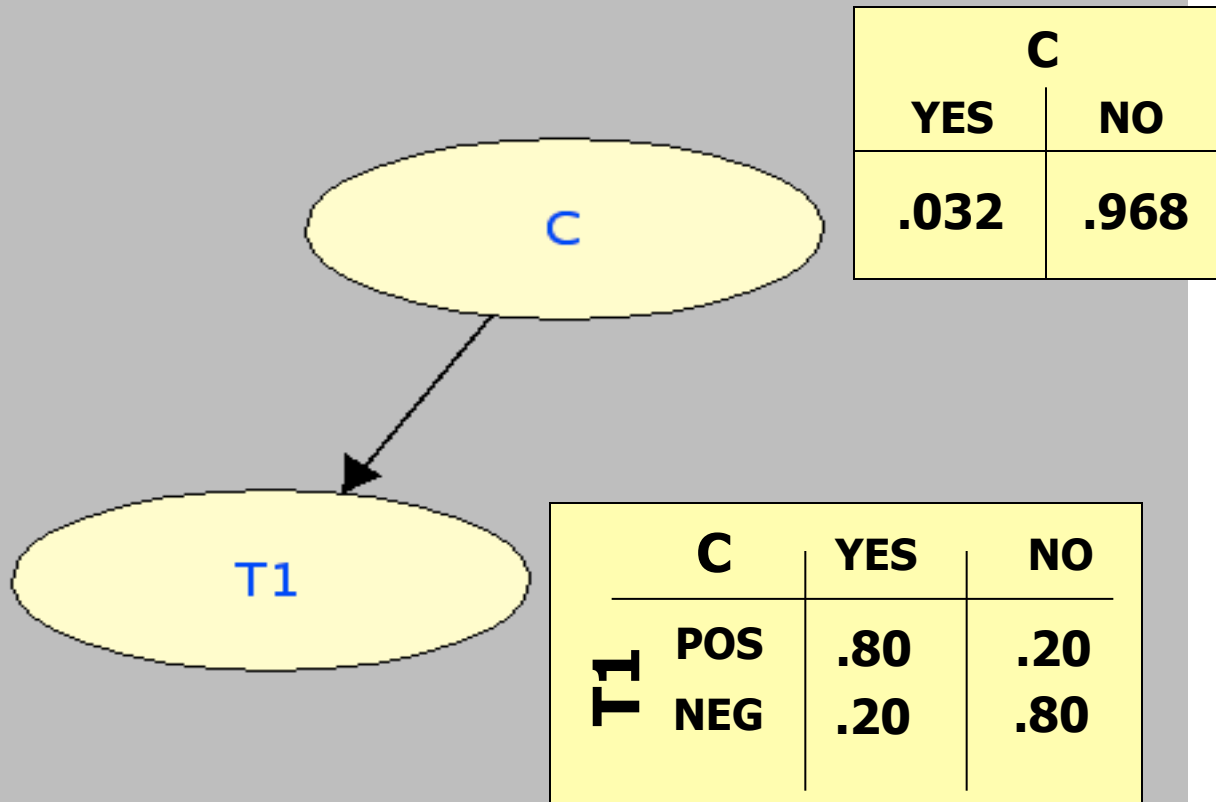
CS161: Introduction to Artificial Intelligence

Adnan Darwiche

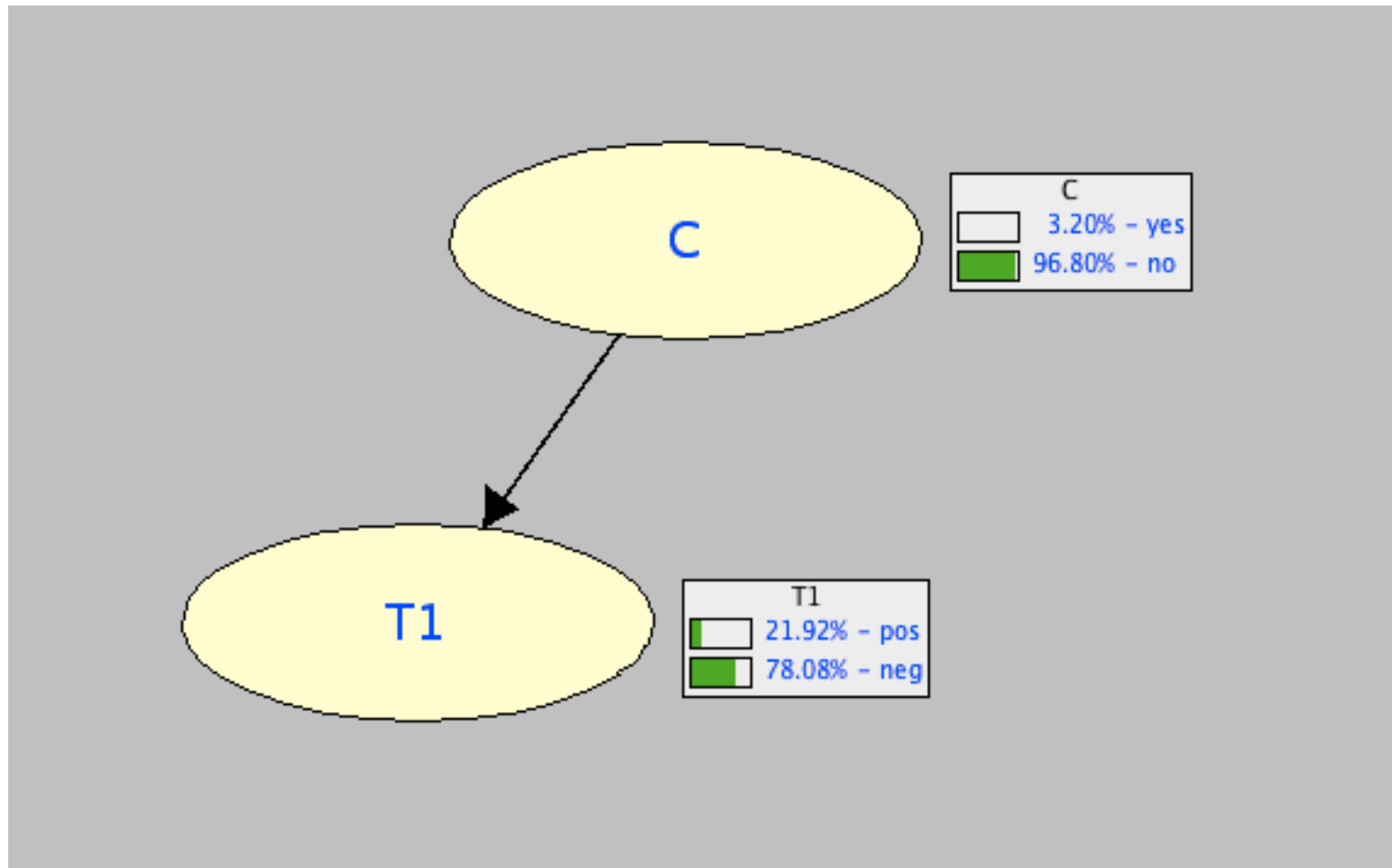
A Simple Problem

- Jack suspects having a medical condition which hits 3.2% of the population
- He takes a test that has a false positive/negative rate of 20%
- The test comes out positive
- What's the probability that Jack has the condition?

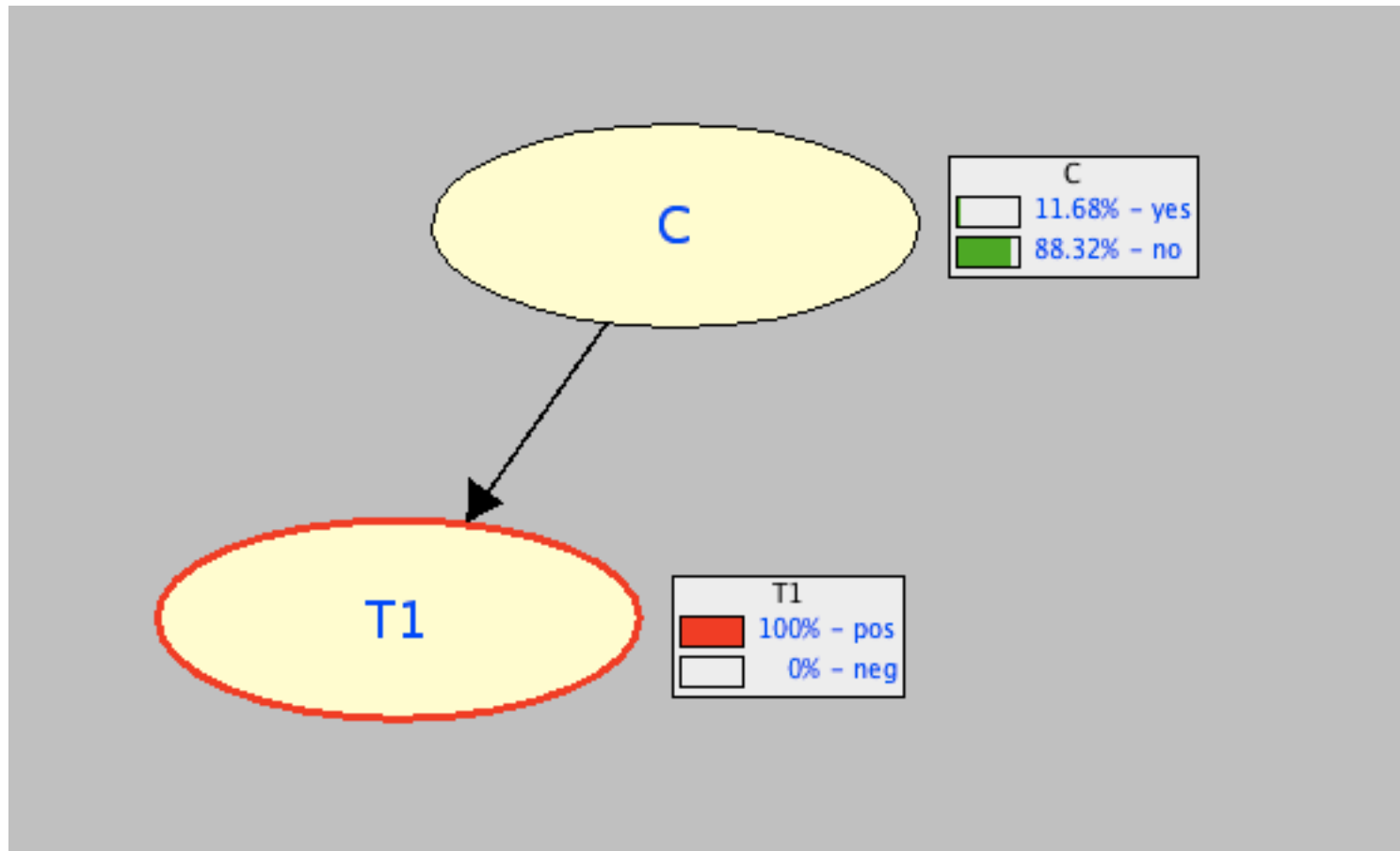
A Simple Network



Prior Marginals



Posterior Marginals



Sensitivity Analysis

Confirm condition with probability greater than 25%

Event

C

Constraint

\geq

.25

= Pr(yes)=0.116...

Start sensitivity an...

Settings Tool
s



Constrain Two Events



Show Table Details

Single parameter suggestions

Parameter	Current value	Suggested value
Pr(C = yes)	0.032	≥ 0.076923
Pr(T1 = pos C = no)	0.2	≤ 0.079339

False positive should be no more than $\sim 7.9\%$

C
11.68% - yes
88.32% - no

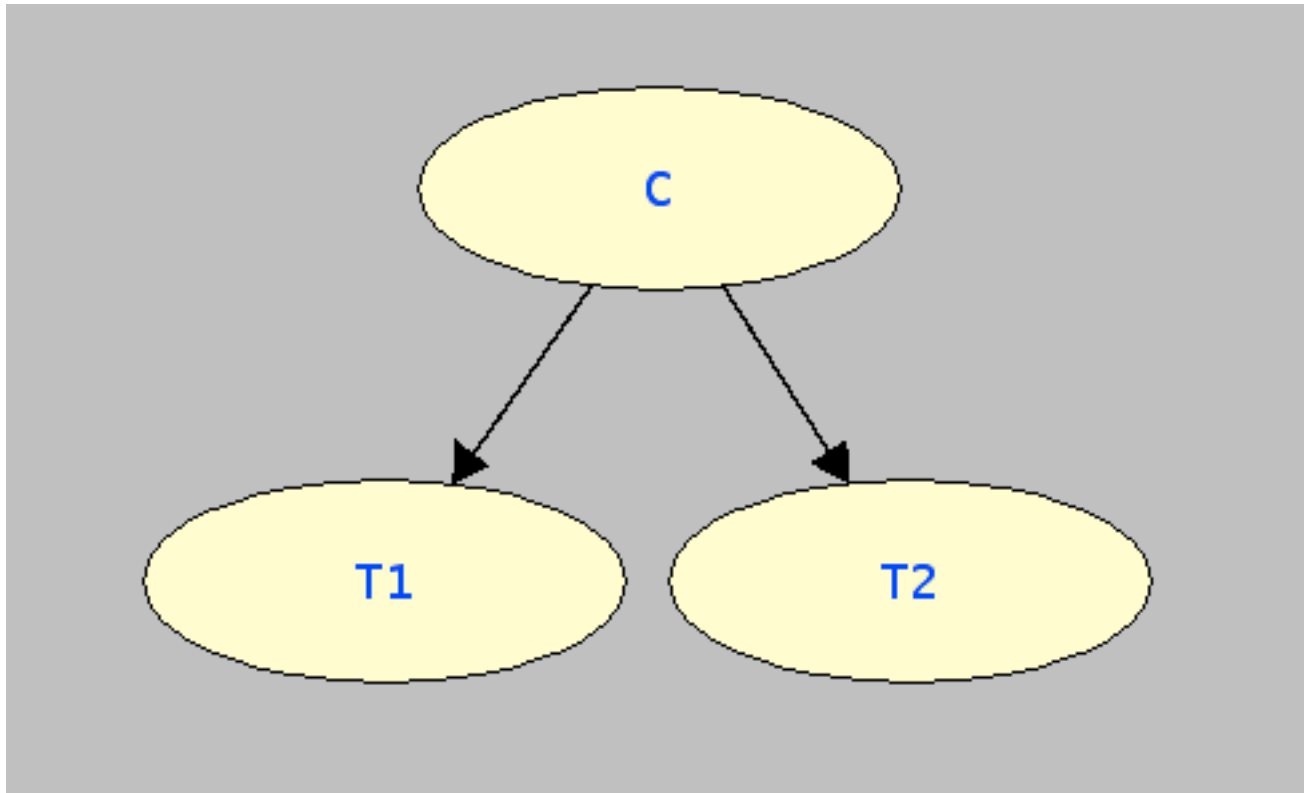
T1
100% - pos
0% - neg

T1

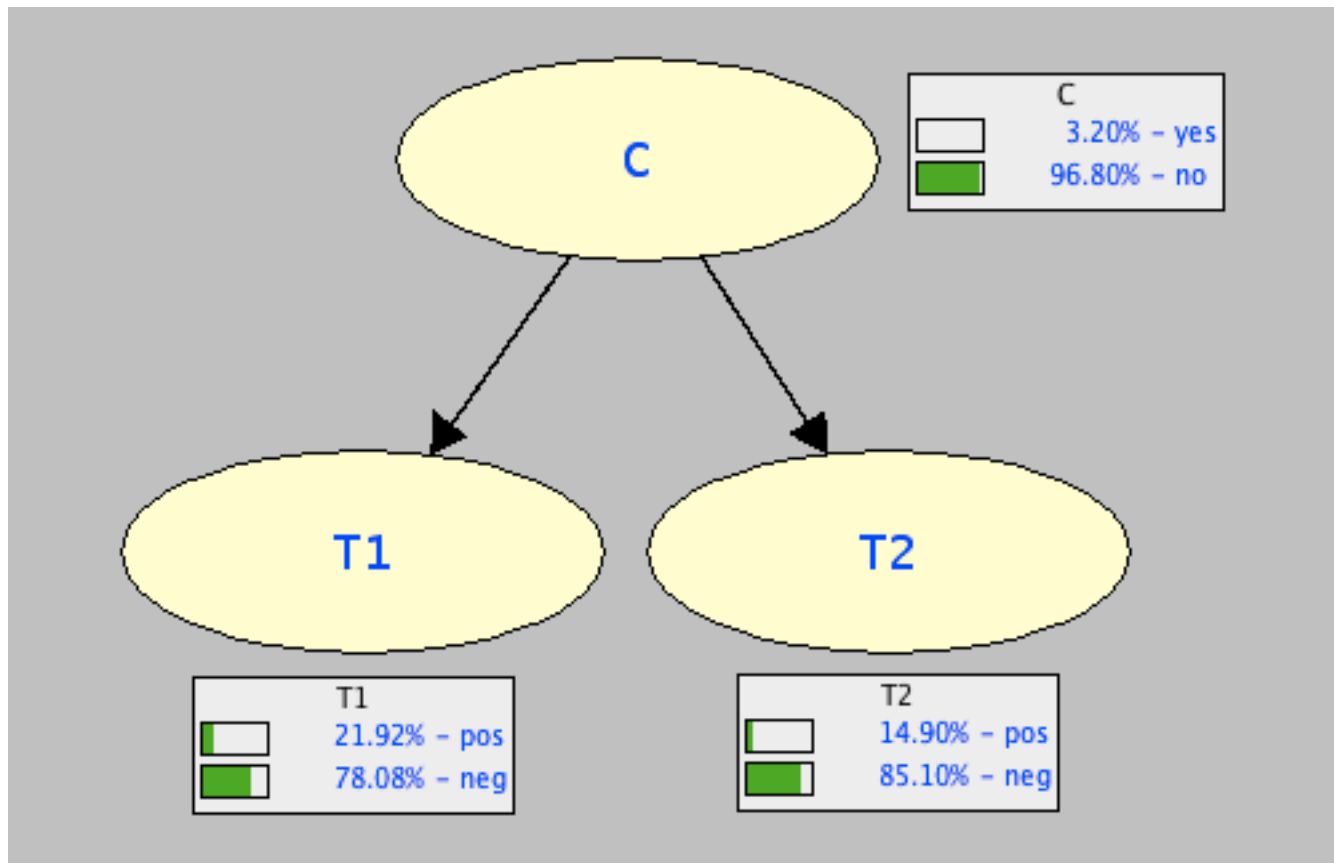
Another test

- Jack took another test that has a false positive/negative rate of 12.5%
- The test also came out positive
- What's the probability that Jack has the condition?

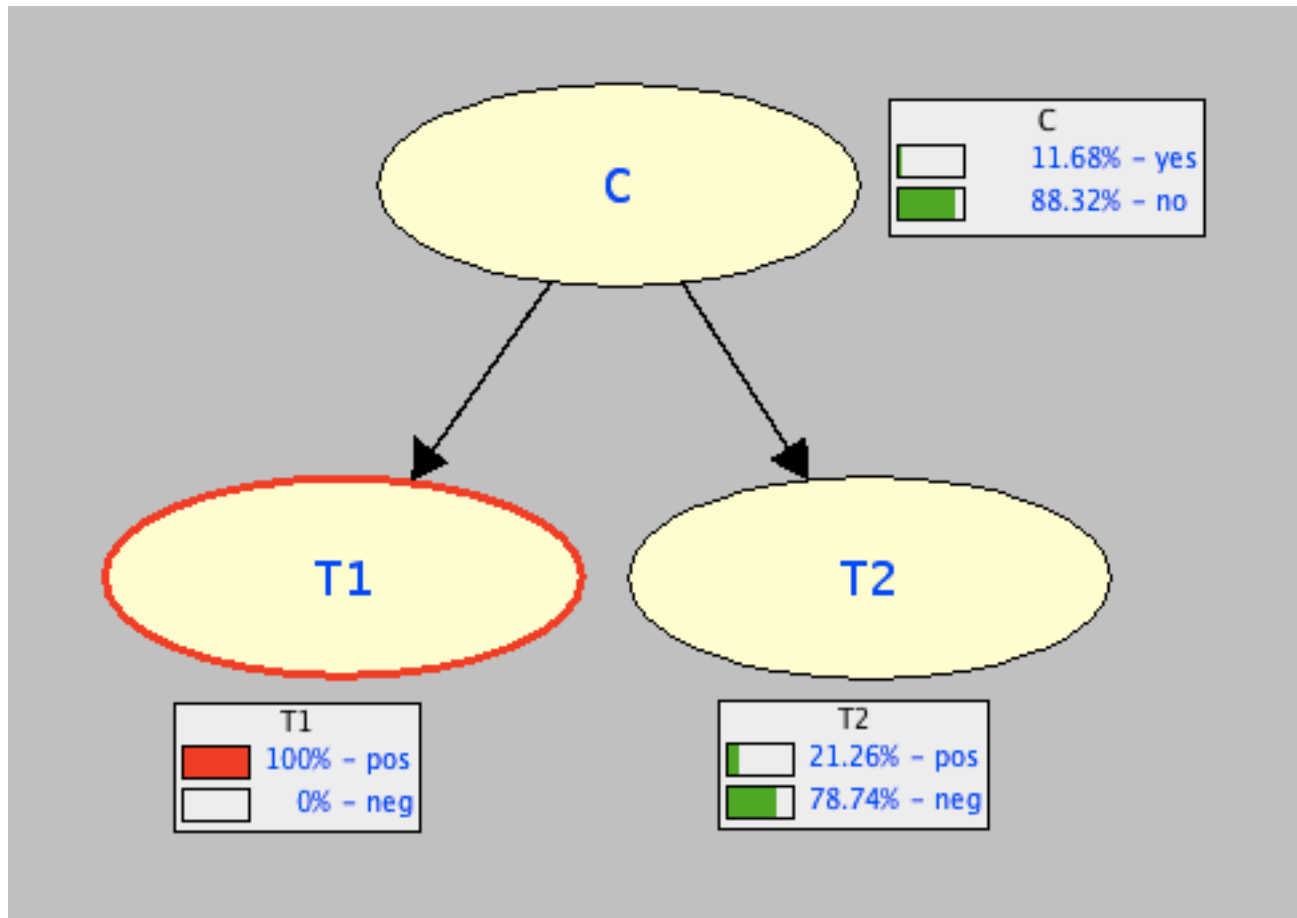
Another test



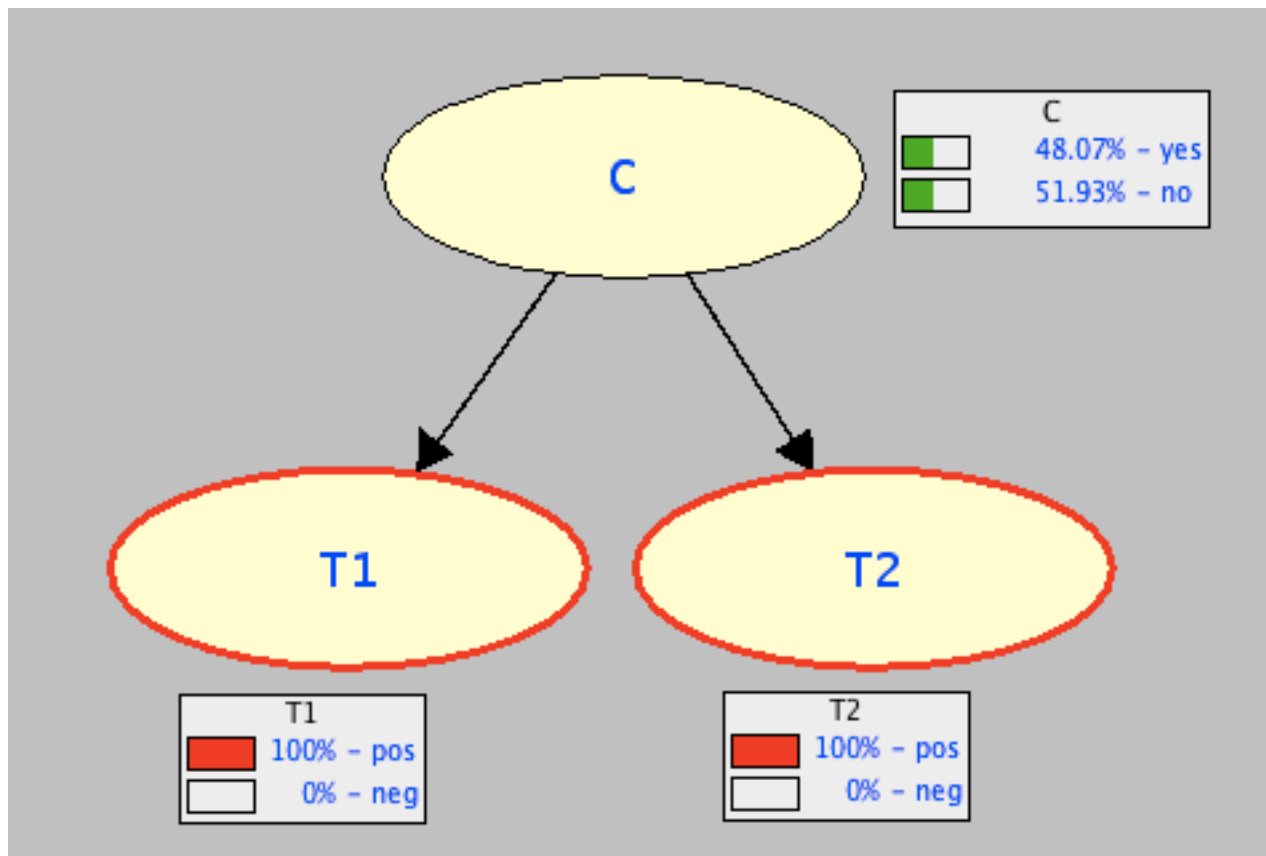
Another test



Another test

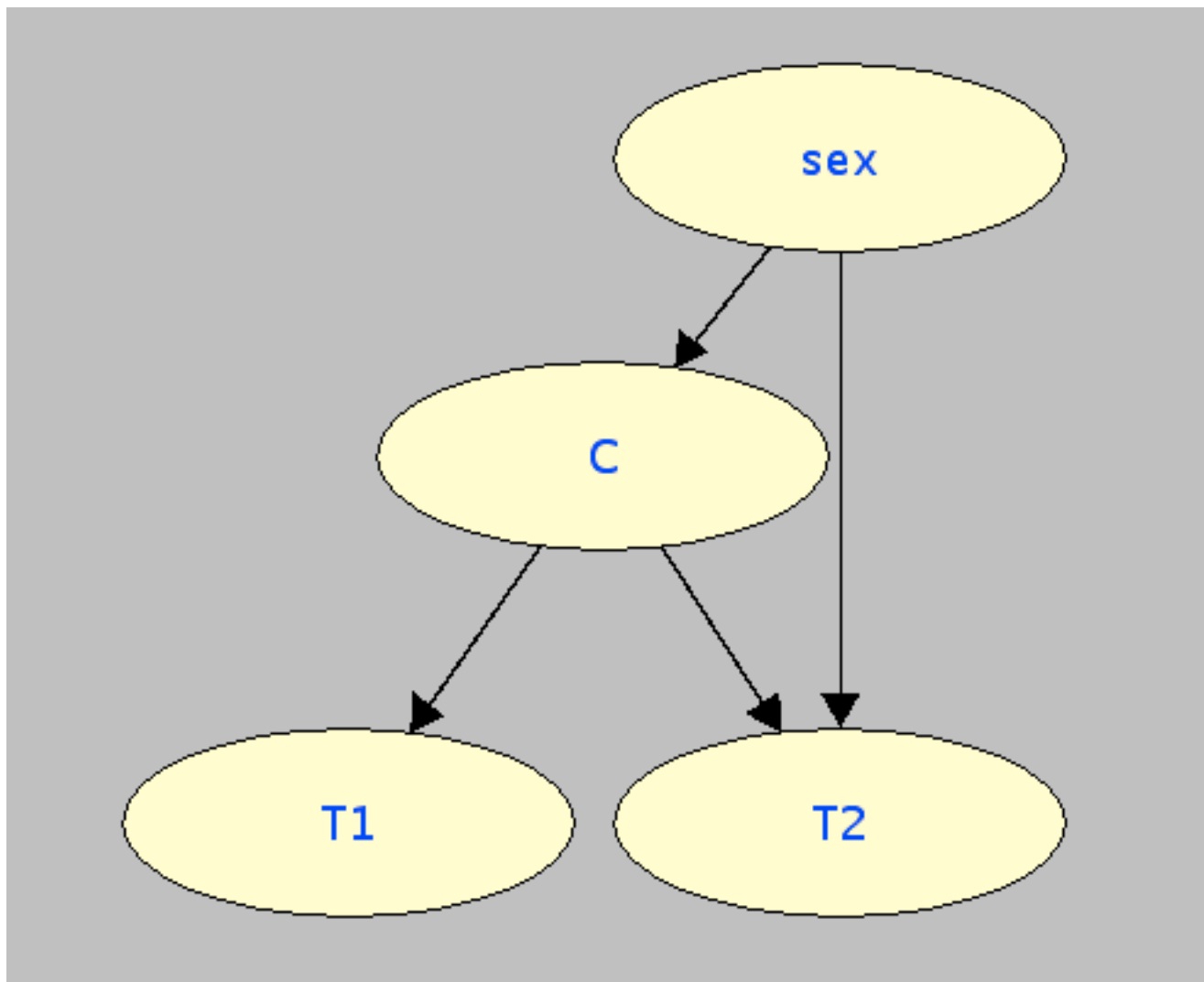


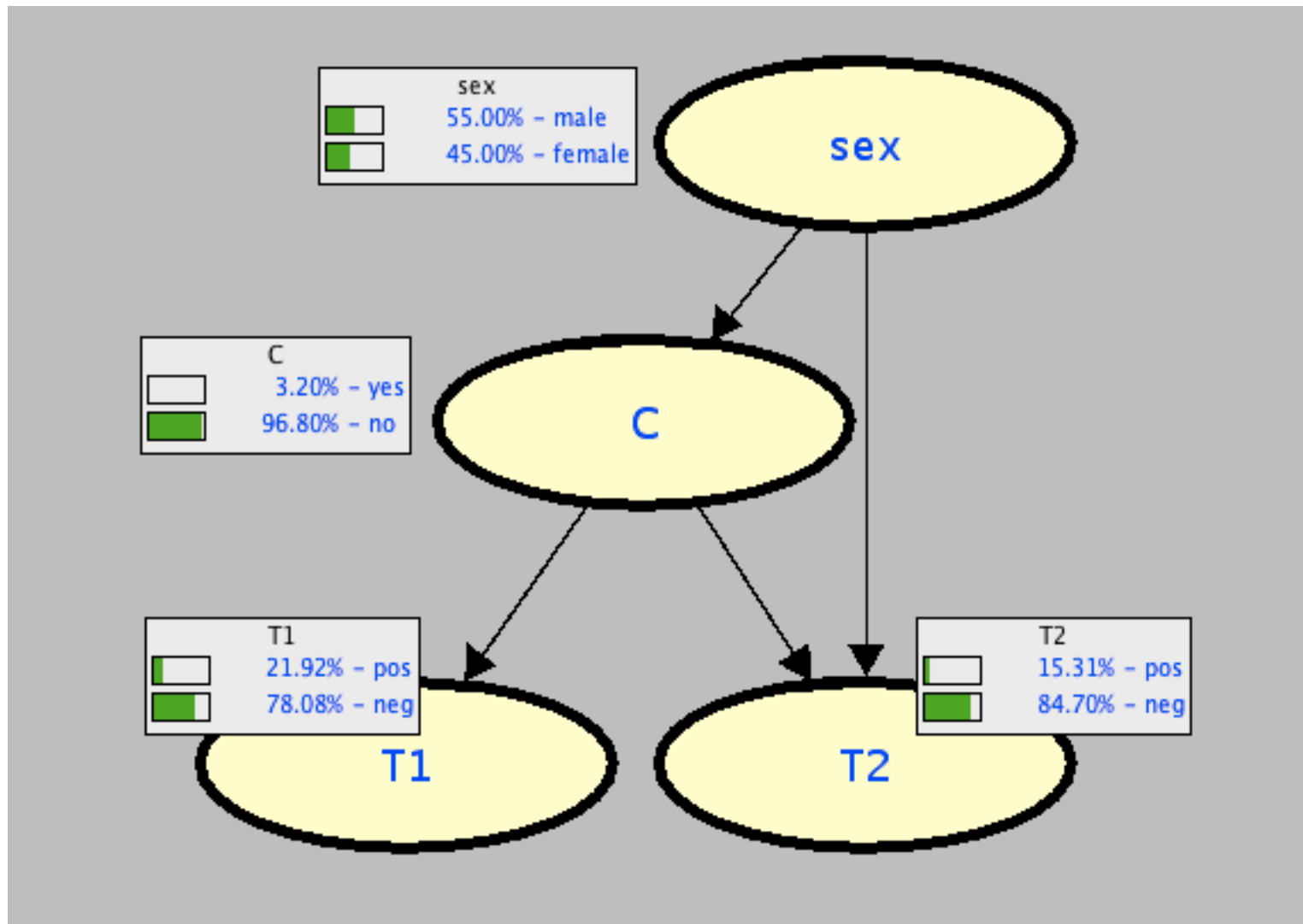
Another test



Introducing Gender

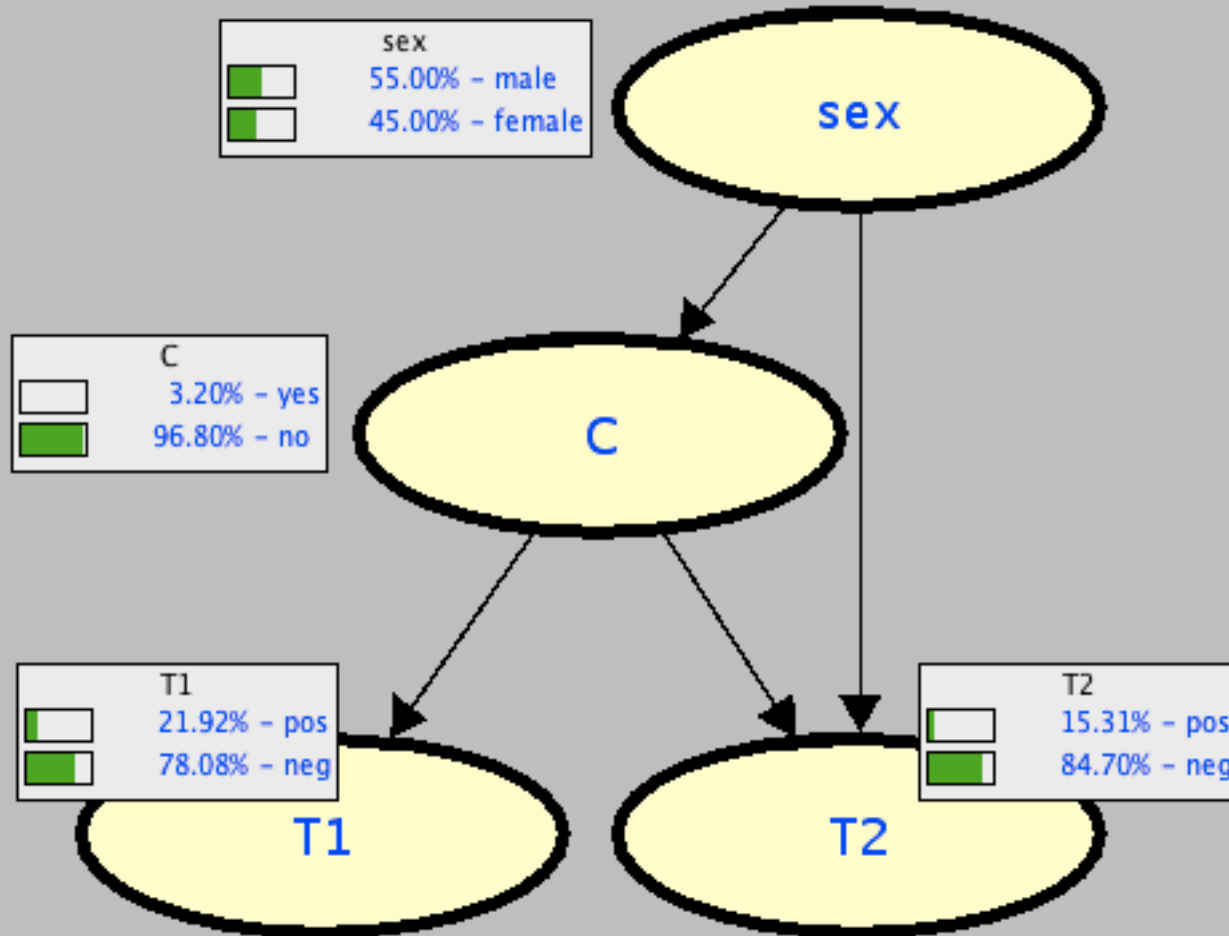
- The condition is more common in males (5%) than females (1%)
- The second test is more effective on females (...)

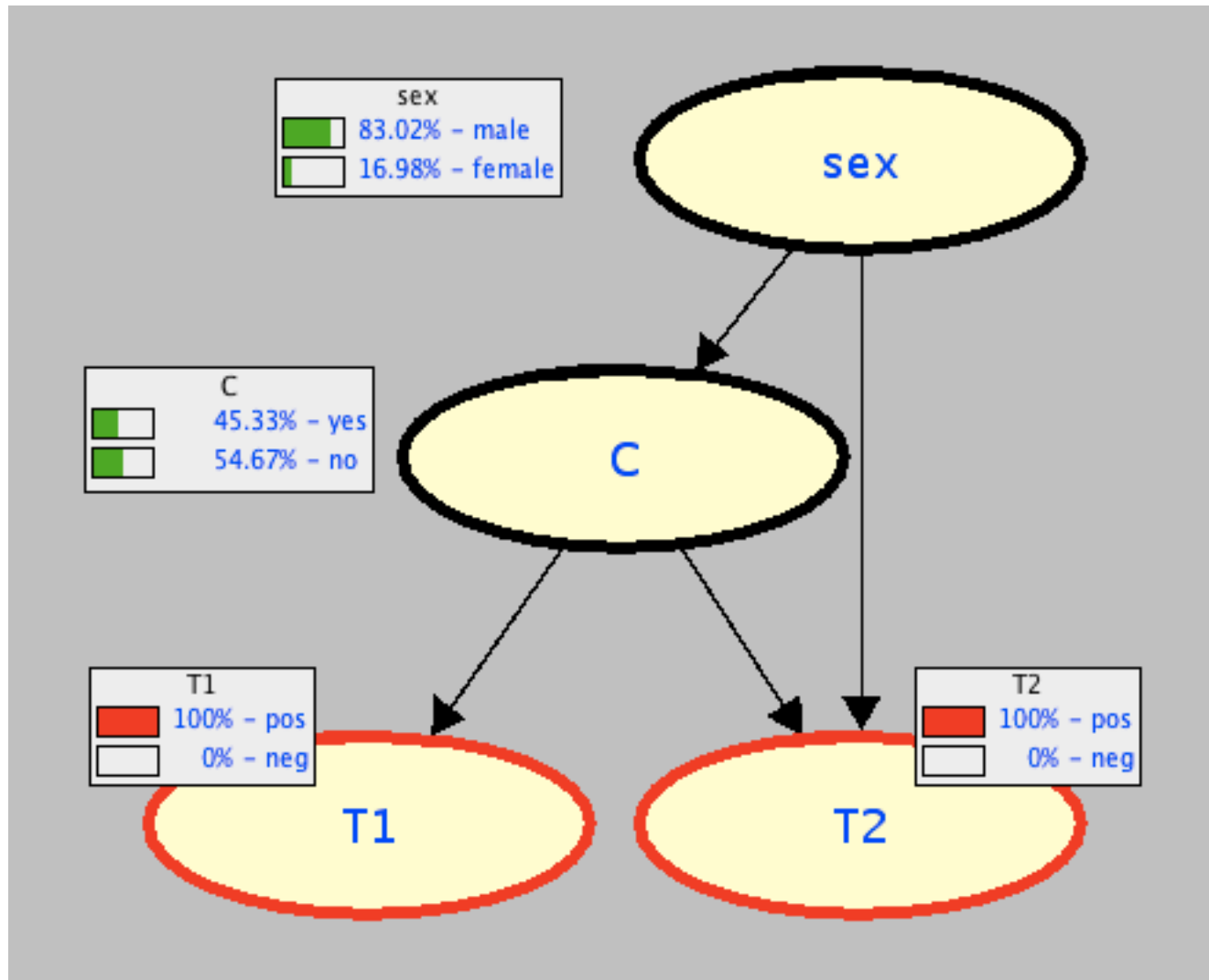


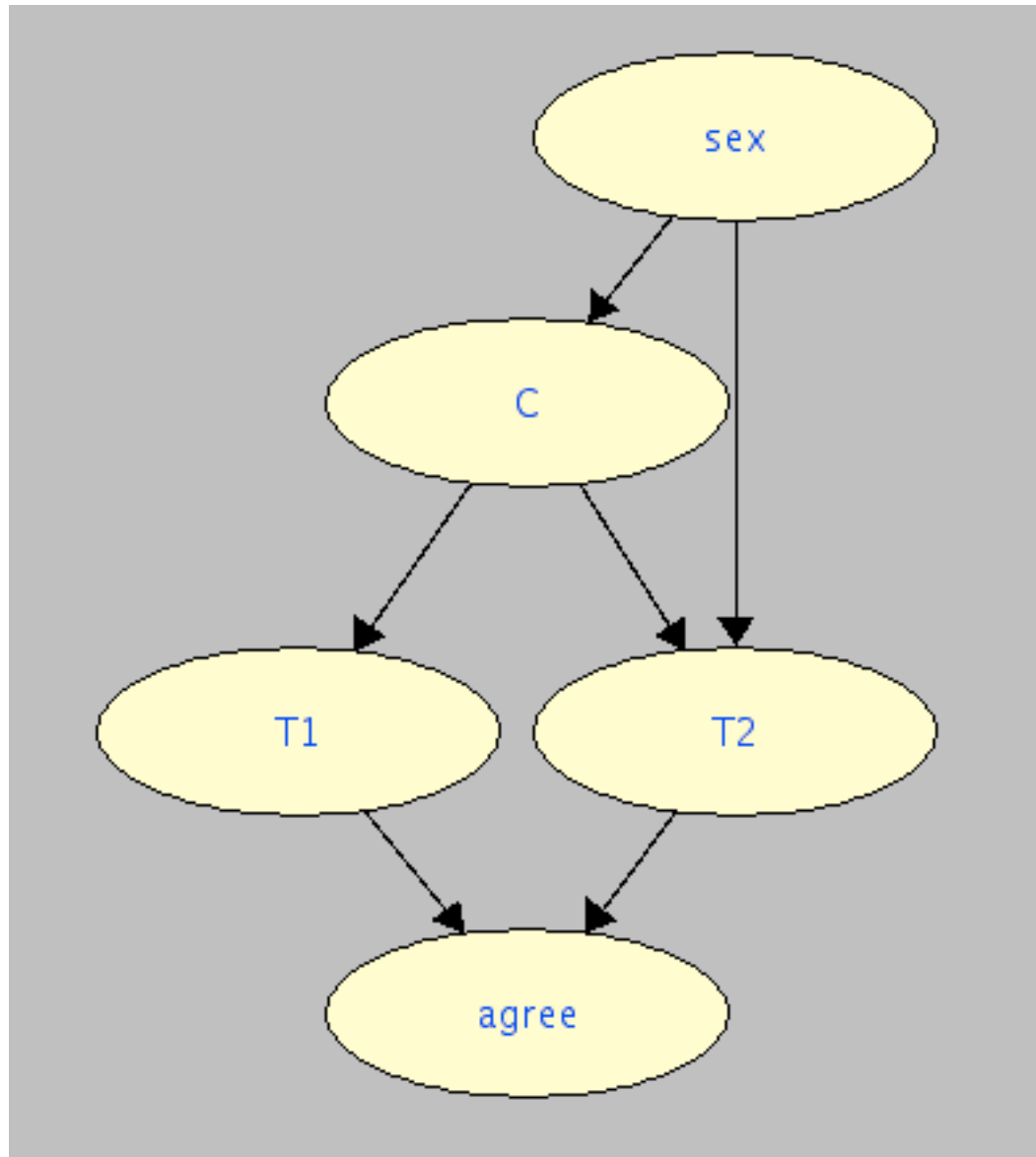


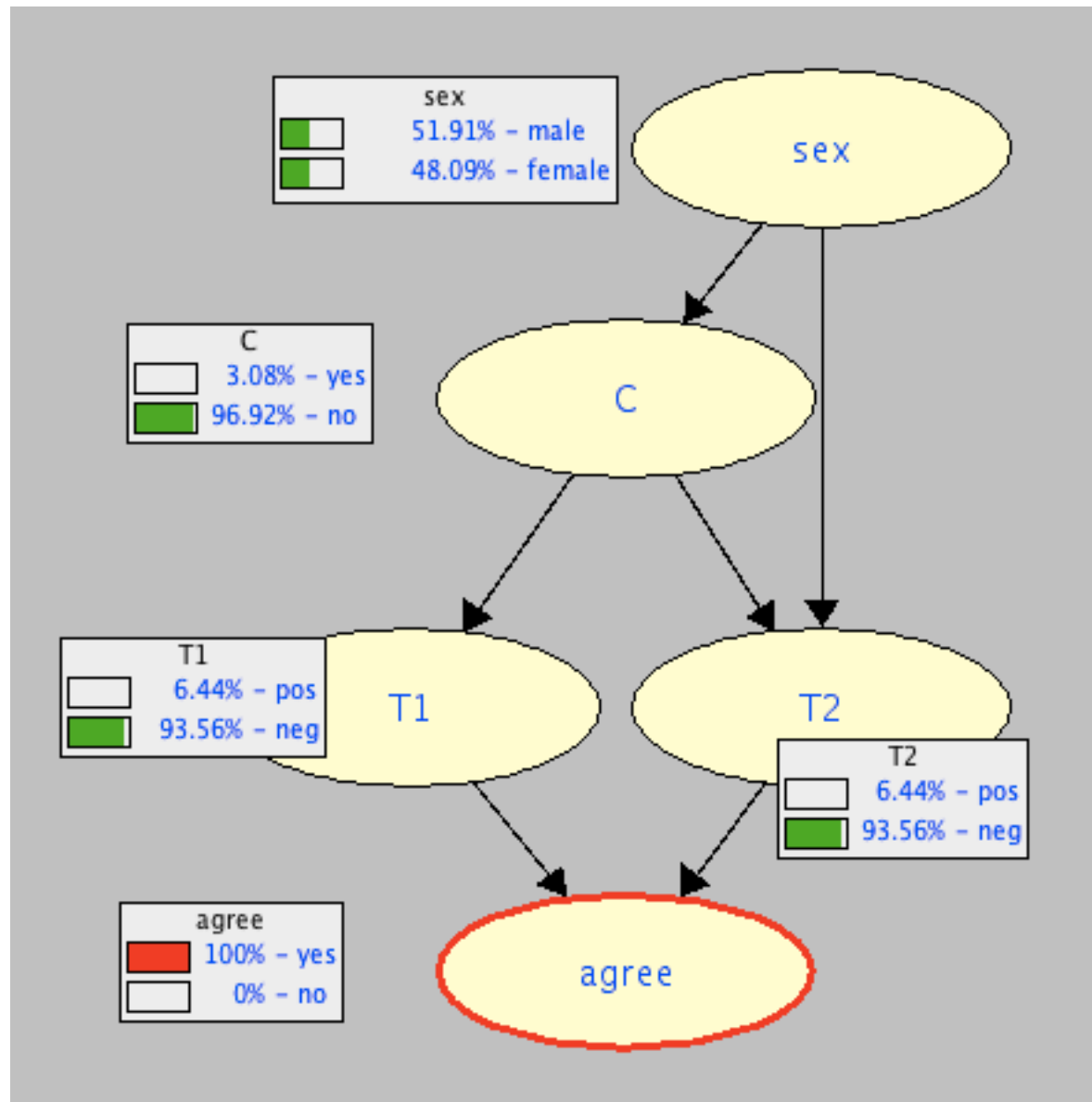
Independencies implied by network:

- Given C, T1 and T2 are independent
- Given C, T1 is independent of Sex

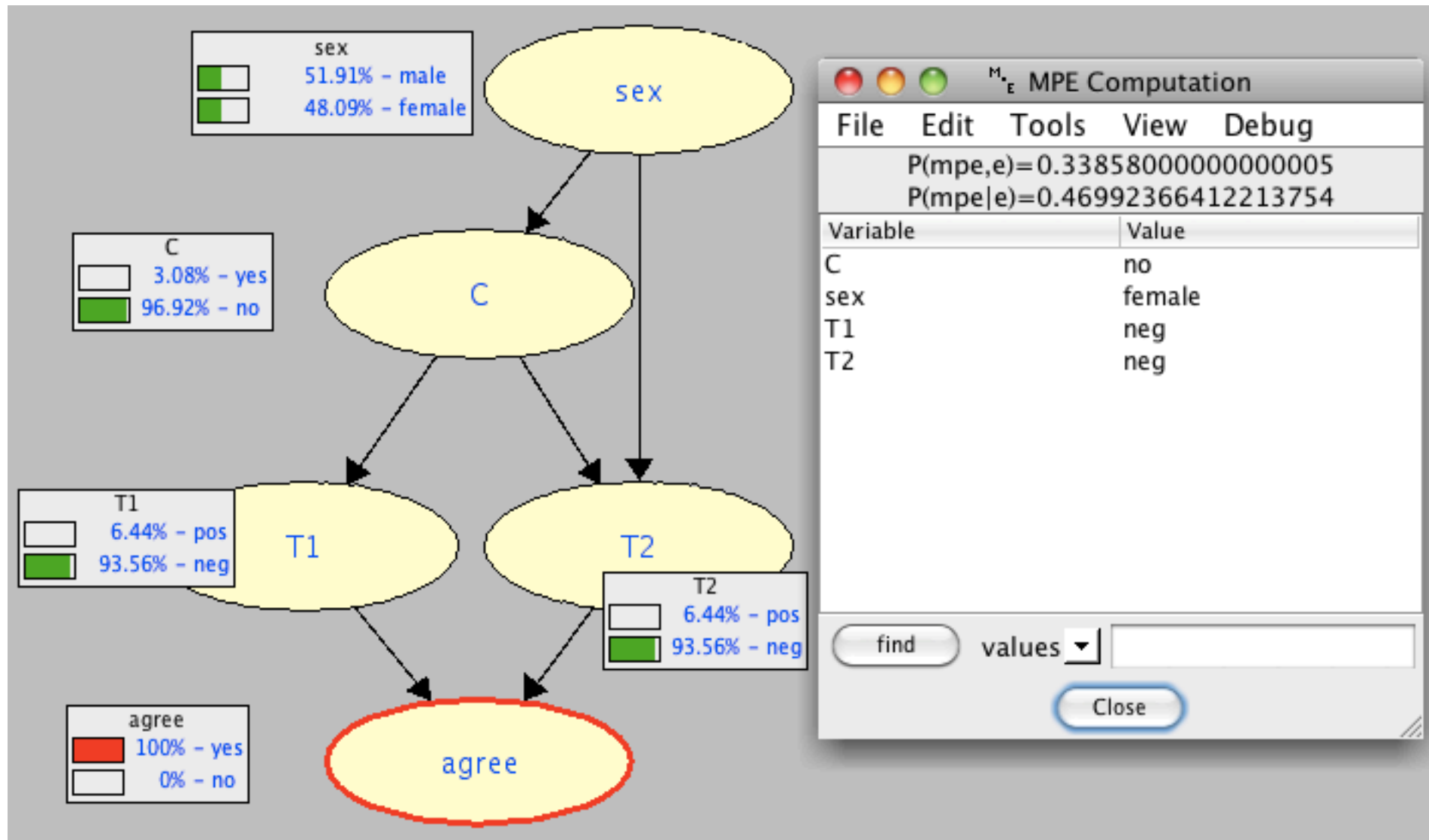




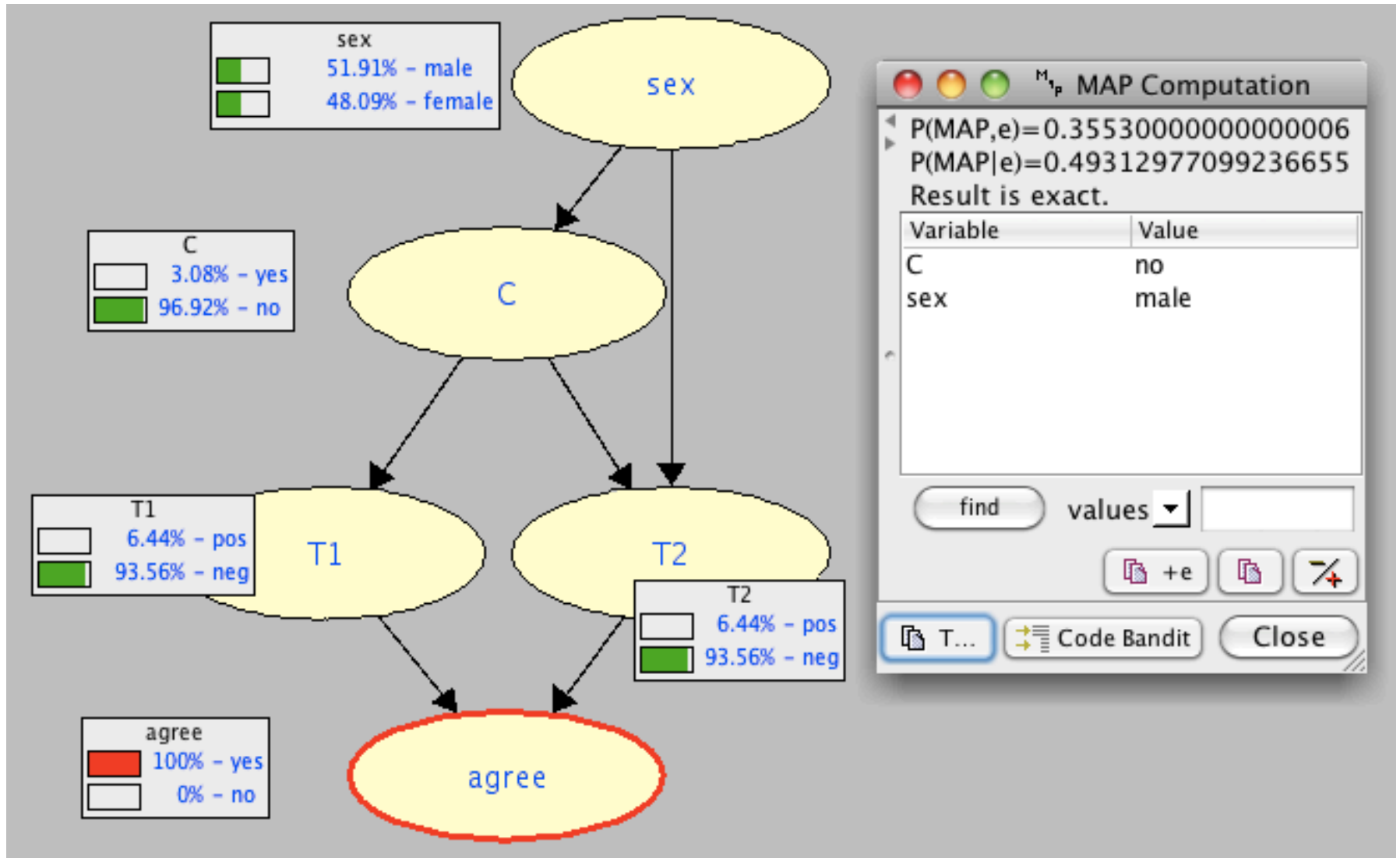




Most Probable Explanation (MPE)



Maximum a Posteriori (MAP)

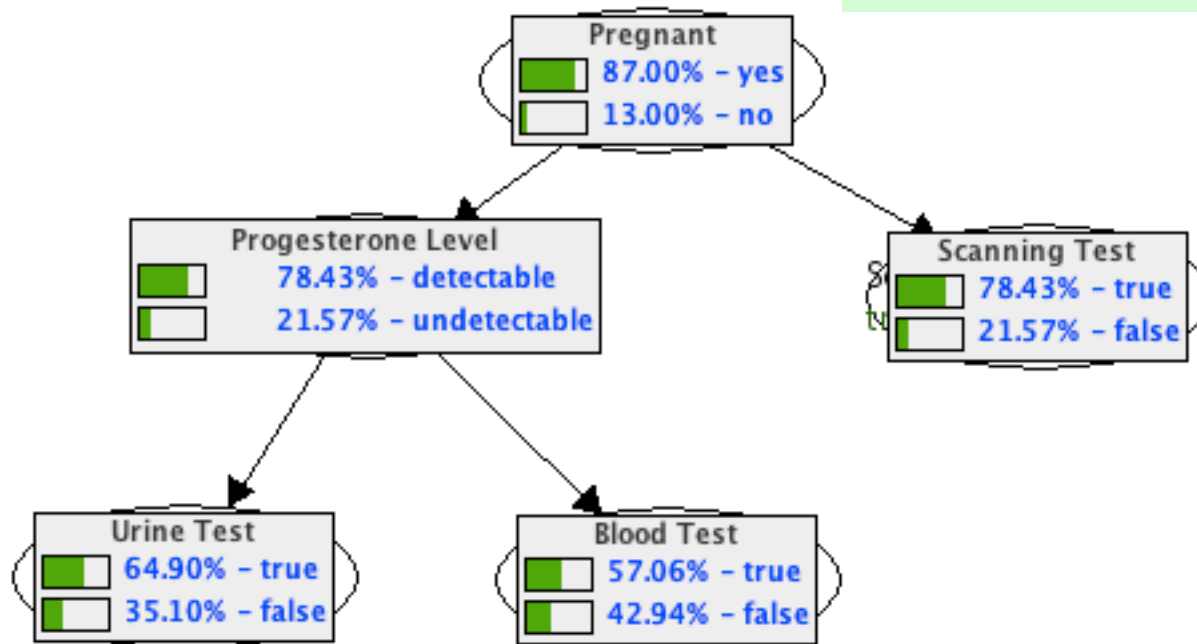


Query Types

- **Prior & Posterior Marginals**
(PP-complete)
- **MPE: Most probable explanations**
(NP-complete)
- **MAP: Maximum a Posteriori Hypothesis**
(NP^{PP}-complete)
- **SDP: Same-Decision Probability**
(PP^{PP}-complete)

Same-Decision Probability

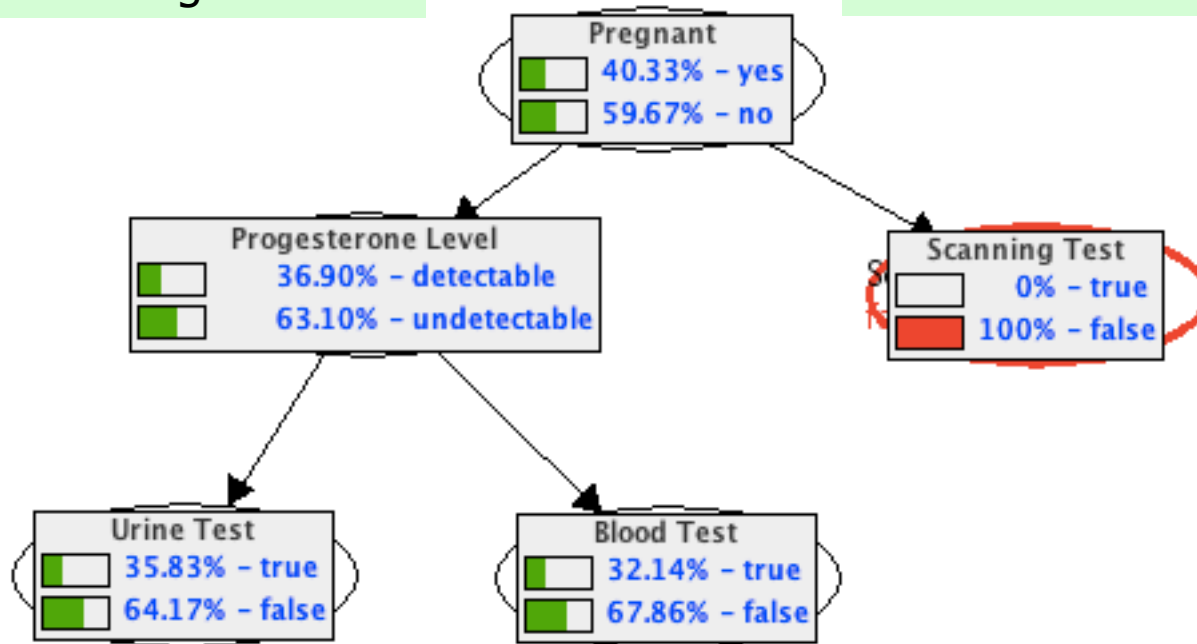
Threshold for Pregnancy is 90%



Same-Decision Probability

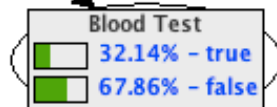
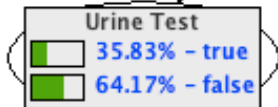
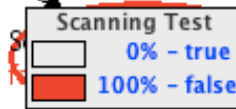
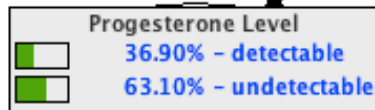
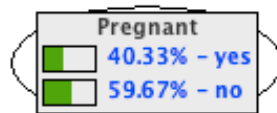
Decision is Pregnant=no

Threshold for Pregnancy is 90%

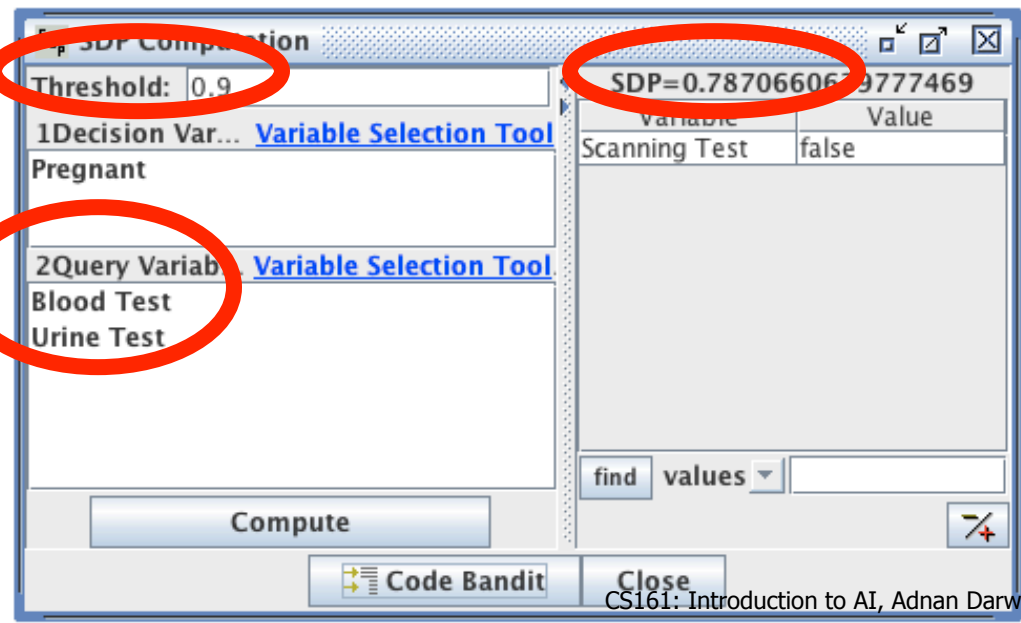


Same-Decision Probability

Decision is
Pregnant=no



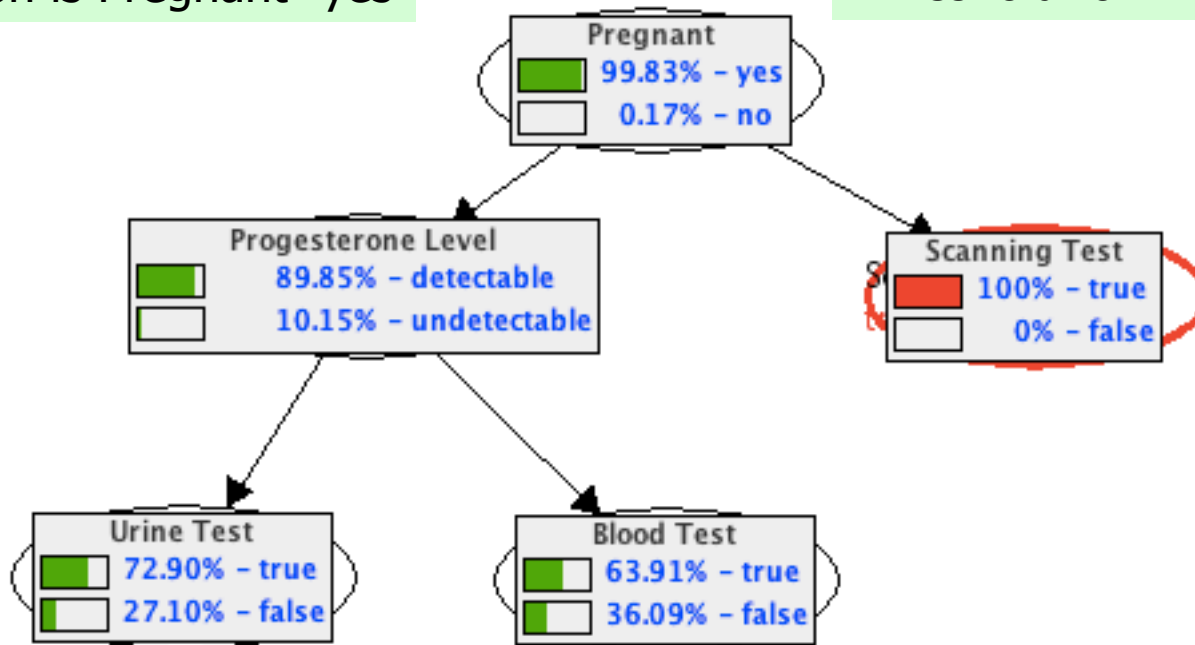
78.7% chance you will still make the same decision after collecting the blood and urine tests.



Same-Decision Probability

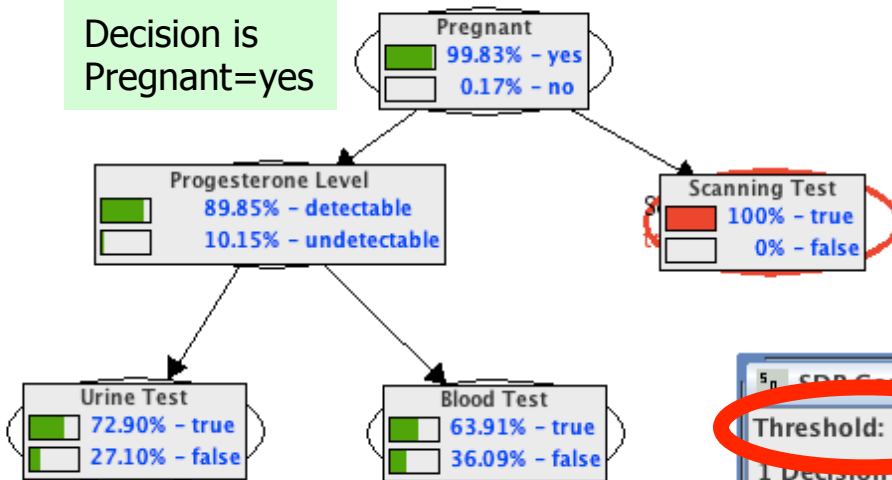
Decision is Pregnant=yes

Threshold for Pregnancy is 90%

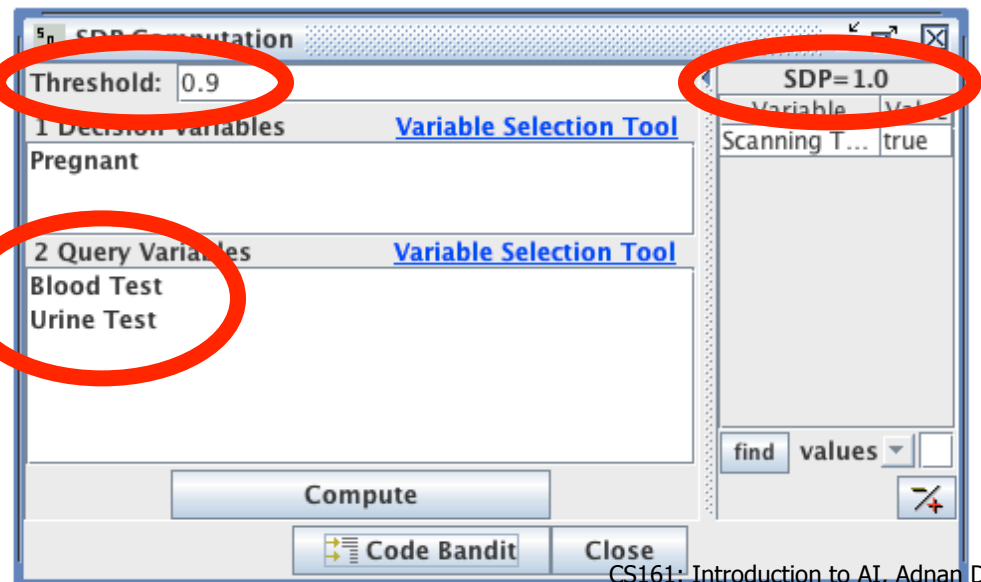


Same-Decision Probability

Decision is
Pregnant=yes



100% chance you will still make the same decision after collecting the blood and urine tests.



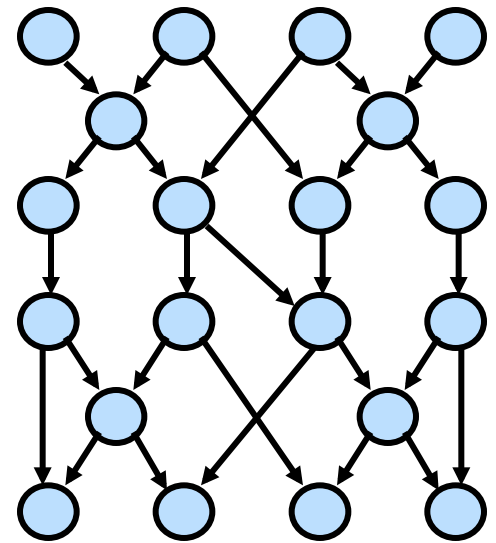
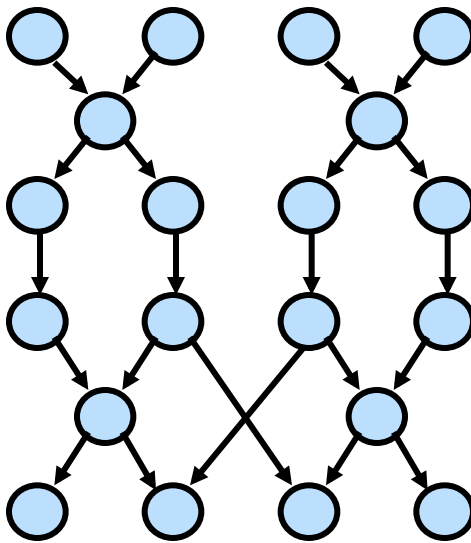
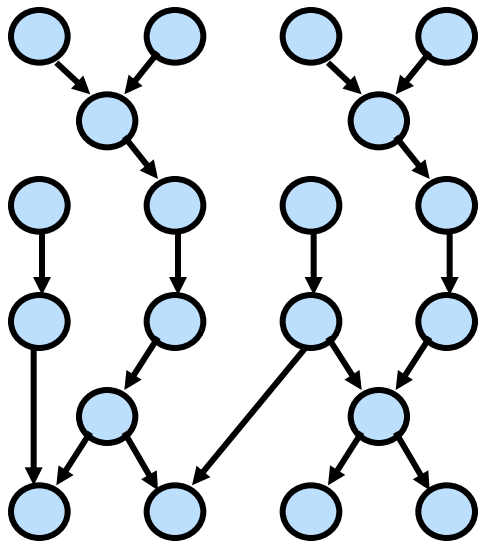
Same-Decision Probability

probability that we would have made the same decision had we known some additional information.

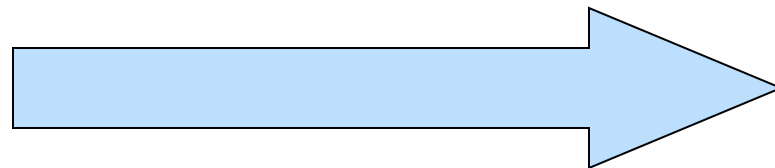
Some Applications

- Evidence: symptoms
Query: **what disease?**
- Evidence: phenotype, genotype
Query: **prone to disease?**
- Evidence: words in a document
Query: **what topics?**
- Evidence: pixels in an image
Query: **how far are objects from camera?**
- Evidence: credit card transaction
Query: **is it the owner?**
- Evidence: Speech
Query: **what words were uttered?**

Model Structure: Treewidth w



$$O(n \exp(w))$$



A. Darwiche



AUTOMATED REASONING GROUP

UNIVERSITY OF CALIFORNIA LOS ANGELES

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SamIam

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Try ACE - a companion system for networks exhibiting local structure: determinism and CSI

SamIam is a comprehensive tool for modeling and reasoning with Bayesian networks, developed in Java by the Automated Reasoning Group of Professor Adnan Darwiche at UCLA.



SamIam includes two main components: a graphical user interface and a reasoning engine. The graphical interface allows users to develop Bayesian network models and to save them in a variety of formats. The reasoning engine supports many tasks including: classical inference; parameter estimation; time-space tradeoffs; sensitivity analysis; and explanation-generation based on MAP and MPE.

AR Group, UCLA



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<http://reasoning.cs.ucla.edu/samiam/>

Constructing Bayesian Networks

- Subjective beliefs
- Learning from data
- Synthesis from other type of knowledge

NASA Ames, Mountain View

NOTES: UNLESS OTHERWISE SPECIFIED

CABINET 1 - POWER GENERATION UNIT

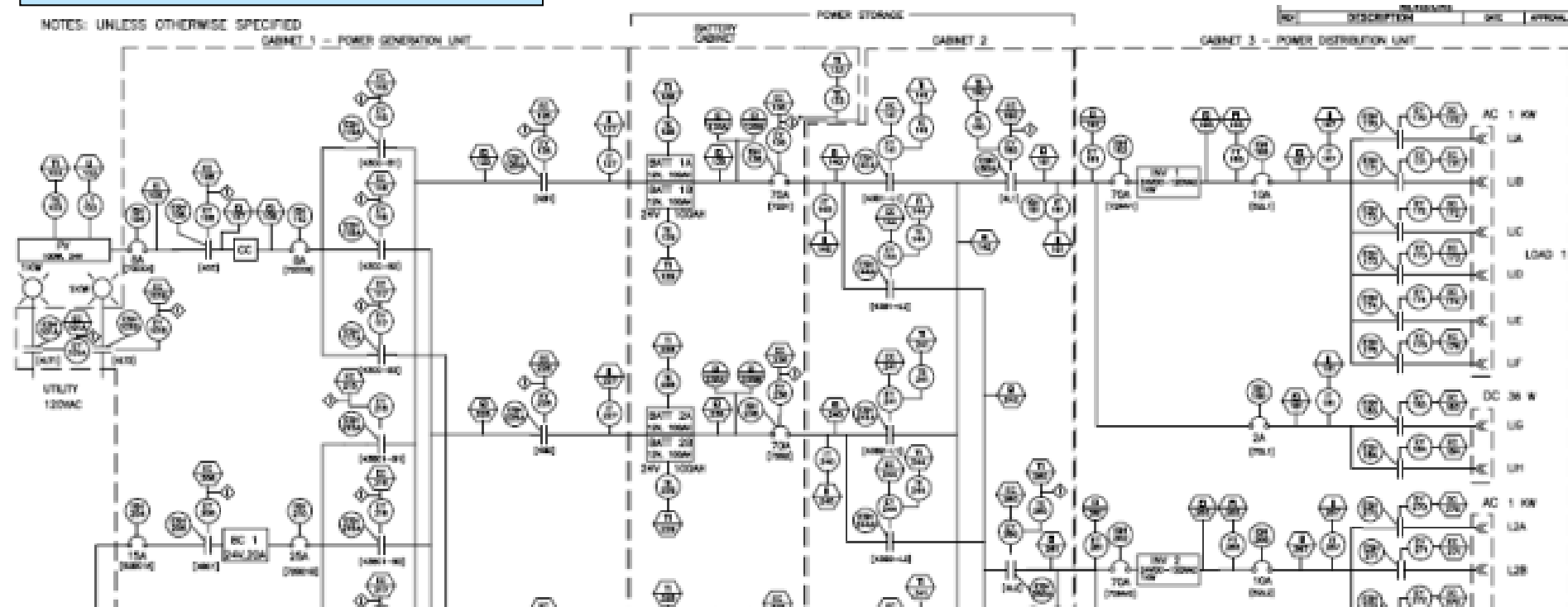
BATTERY CABINET

POWER STORAGE

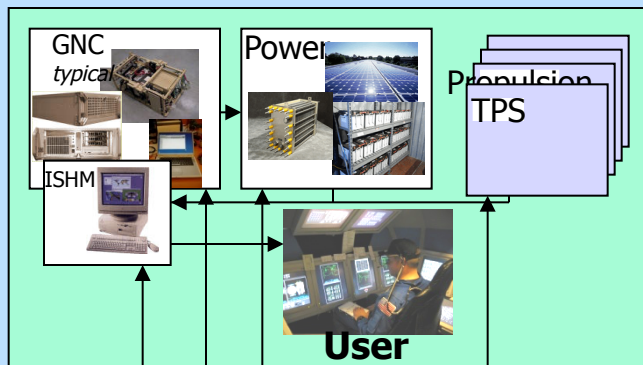
CABINET 2

CABINET 3 - POWER DISTRIBUTION UNIT

REV: DISCREPTION: DATE: APPROVAL:



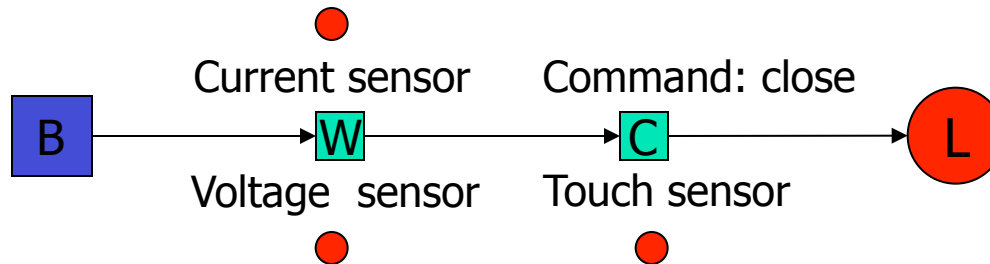
ADAPT Testbed



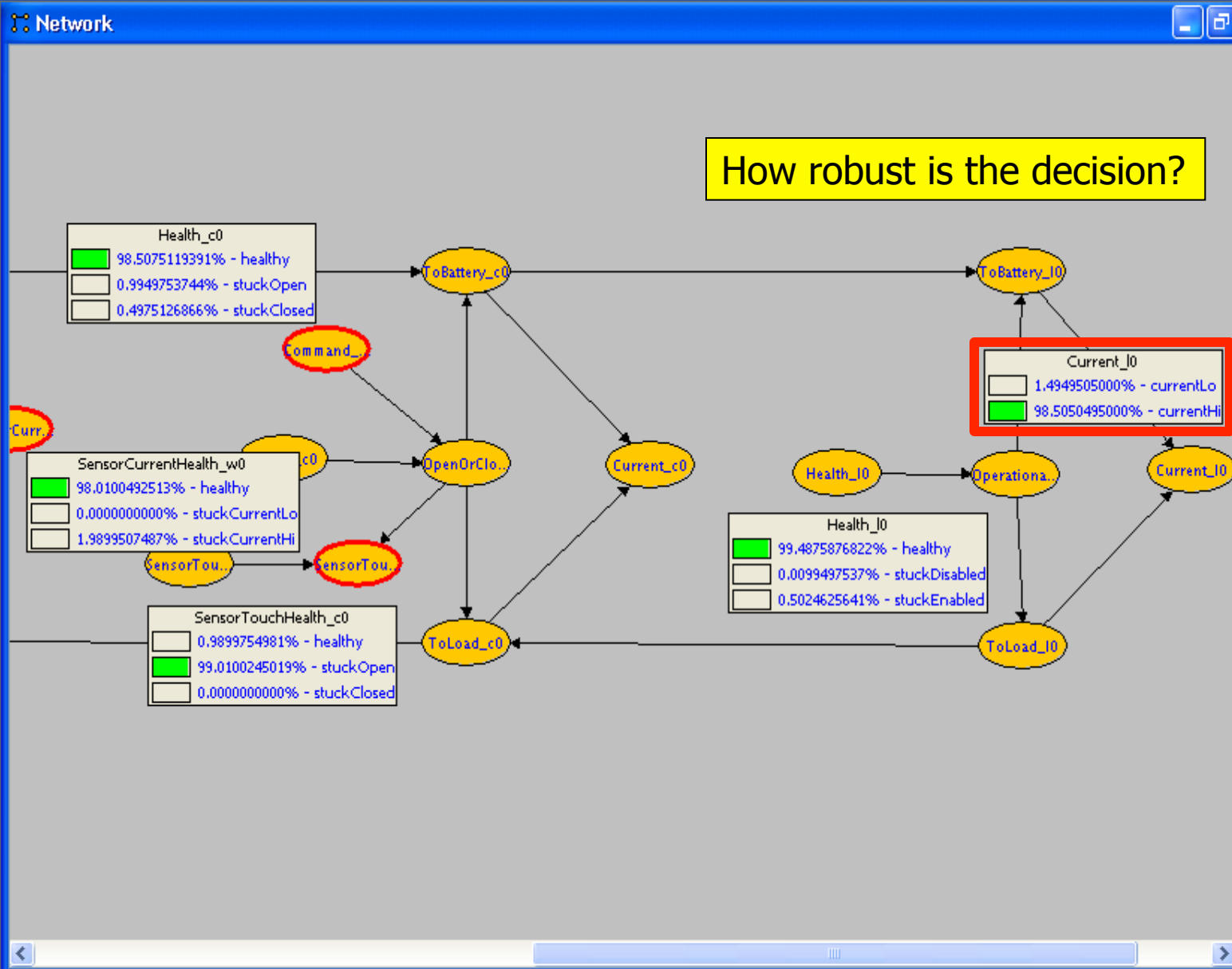
Antagonist Observer



Monitoring and Diagnosis



- adaptkind
- sensor
 - SensorCurrent_w0
 - readCurrentLo
 - readCurrentHi
 - SensorTouch_c0
 - readOpen
 - readClosed
 - SensorVoltage_w0
 - readVoltageLo
 - readVoltageHi
 - command
 - Command_c0
 - cmdOpen
 - cmdClose
 - health
 - Health_b0
 - Health_c0
 - Health_I0
 - SensorCurrentHealth_y
 - SensorTouchHealth_c0
 - SensorVoltageHealth_y
 - current
 - Current_b0
 - Current_c0
 - Current_I0
 - Current_w0
 - aux
 - OpenOrClosed_c0
 - OpenOrClosed_w0
 - Operational_b0
 - Operational_I0
 - ToBattery_b0
 - ToBattery_c0
 - ToBattery_I0
 - ToBattery_w0
 - ToLoad_b0



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Results in DX Competition

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Diagnostic Inference using Probabilistic Computation



Results in DX Competition

We have participated in a diagnostic challenge that was organized as part of the 20th International Workshop on Principles of Diagnosis (DX-09); see <http://www.isy.liu.se/dx09/> for information about the workshop. The challenge was organized in two tracks, an Industrial Track and a Synthetic Track, and we participated in the Industrial Track featuring the ADAPT electrical power system (EPS) testbed.

Two categories of scenarios were featured in the Industrial Track, namely Tier 1 scenarios (a subset of ADAPT) and Tier 2 scenarios (complete ADAPT). Tier 1 scenarios were nominal or contained one fault. Tier 2 scenarios were nominal or contained single, double, or triple faults.

Faults inserted into ADAPT had the following characteristics:

- Faults were injected simultaneously or sequentially
- Fault types were parametric (change in continuous parameter value) or discrete (change in system mode)
- Faults were abrupt (immediate onset) and permanent
- Faults were to components as well as sensors

Using techniques discussed here, our ProADAPT team obtained the highest scores in both Tier 1 (among 9 international competitors) and Tier 2 (among 6 international competitors) of the Industrial Track of the DX'09 Diagnostic Challenge Competition. One key component of our ProDiagnose algorithm was the use of a Bayesian network model of ADAPT, which was compiled into an arithmetic circuit, which was then used for on-line diagnosis. For further information on the compilation to arithmetic circuits, please see <http://reasoning.cs.ucla.edu/>. For a high-level discussion of our approach, please see: <http://ti.arc.nasa.gov/project/pca/>.