Architecting for Causal Intelligence at Nanoscale

Supplementary Document

(A)
$$\lambda_{X}(A) = \lambda_{Y}(X) = \lambda_{Y$$

(B)

Bayesian Inference Operations using Pearl's Belief Propagation Algorithm	
Likelihood Estimation Prior Estimation	$\lambda(X) = \lambda_Y(X) * \lambda_Z(X)$ $\pi(X) = \pi_X(A) \otimes CPT(X A)$
Belief Update	$BEL(X) = \alpha \pi(X) * \lambda(X)$
Support Messages: Diagnostic Support to Parent	$\lambda_X(A) = CPT(X A) \otimes \lambda(X),$
Prior Support to Child Nodes	$\pi_{Y}(X) = \alpha \pi(X) * \lambda_{z}(X);$ $\pi_{Z}(X) = \alpha \pi(X) * \lambda_{Y}(X).$

Fig. S- 1. (A) Part of a BN with showing node X with parent A and child nodes Y, Z. All nodes have four states in this example. Each node maintains likelihood vector (λ) , prior vector (π) , belief vector (BEL), and conditional probability table (CPT). The CPT information and messages from child/parent nodes based on observed evidence are used to calculate λ , π , and BEL vectors during Bayesian inference. (B) Overview of computations involved in each node as per Belief propagation algorithm. Here elements in bold typeface indicate vectors as mentioned in (A). The symbol * indicates element-wise multiplication of vectors, which can be implemented using Multiplication Composer circuits. The symbol \otimes indicates vector multiplication that has a sum-of-products form. These operations are implemented using Add-Multiply Composer circuits.

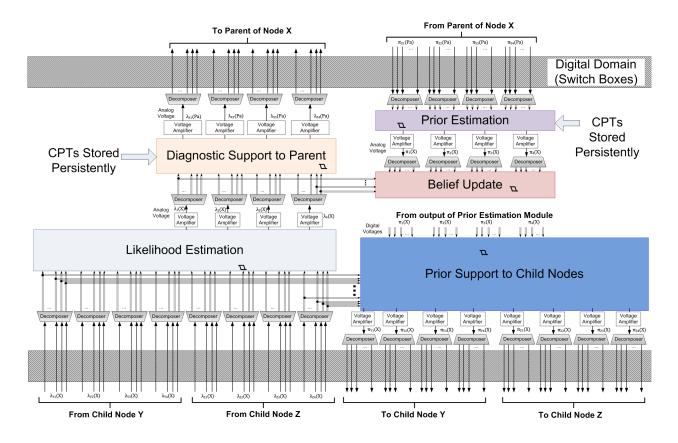


Fig. S- 2. Detailed schematic of the Bayesian Cell showing various modules involved in inference operation. They use Multiplication Composers and Add-Multiply Composers, cascaded through Decomposers [8]. The likelihood, prior and belief vectors, and conditional probability tables (CPTs) are stored persistently in each module.

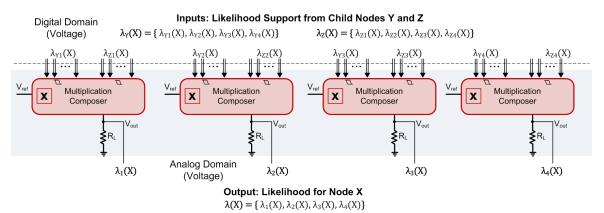
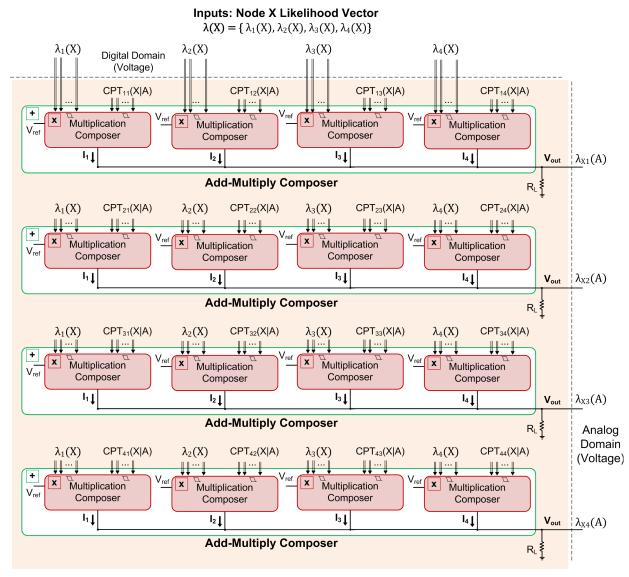


Fig. S- 3. Implementation of Likelihood Estimation operation using Multiplication Composers for a node X, supporting up to four states. Inputs are likelihood support message vectors $\lambda_{\mathbf{Y}}(\mathbf{X})$ and $\lambda_{\mathbf{Z}}(\mathbf{X})$ from child nodes Y and Z respectively. Each vector has four elements corresponding to each state of node X. Belief Update and Prior Support modules use similar implementation.



Output: Diagnostic Support to Parent A for Node X $\lambda_X(A) = \{ \lambda_{X1}(A), \lambda_{X2}(A), \lambda_{X3}(A), \lambda_{X4}(A) \}$

Fig. S- 4. Implementation of Diagnostic Support module using Add-Multiply Composers for a node X. The input is a likelihood vector $\lambda(X)$ for node X. The output is a diagnostic support message vector $\lambda_X(A)$ to be propagated to parent node A. Each vector has four elements corresponding to each state of node X and A. The conditional probability table (CPT) elements are stored persistently in the Composers, obviating the need to interface with external memory. Prior Estimation module similarly uses Add-Multiply Composers.

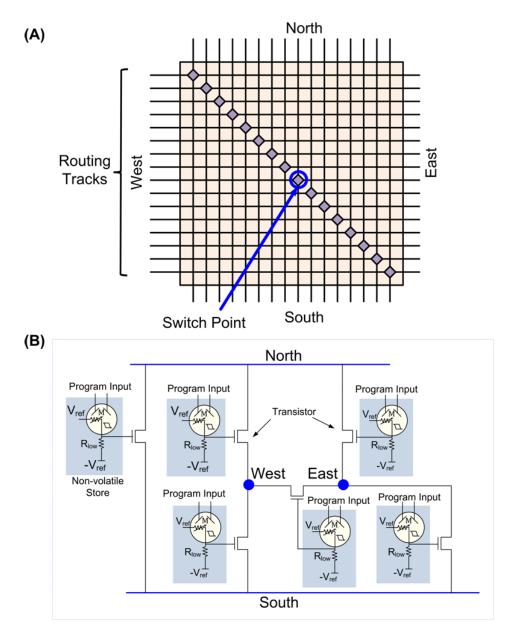


Fig. S- 5. (A) Switch Box schematic: It is a collection of *switch-points*, where each switch-point is capable of routing an incoming signal from one direction to any of three outgoing directions. (B) Switch-point schematic for programmable routing. It consists of six non-volatile S-MTJs gating pass transistors (Metal Oxide Semiconductor Field Effect Transistors - MOSFETs), to be able to activate/deactivate connections between the four directions shown. S-MTJ storing a high resistance deactivates a link, while a low resistance state activates the link allowing signal propagation.