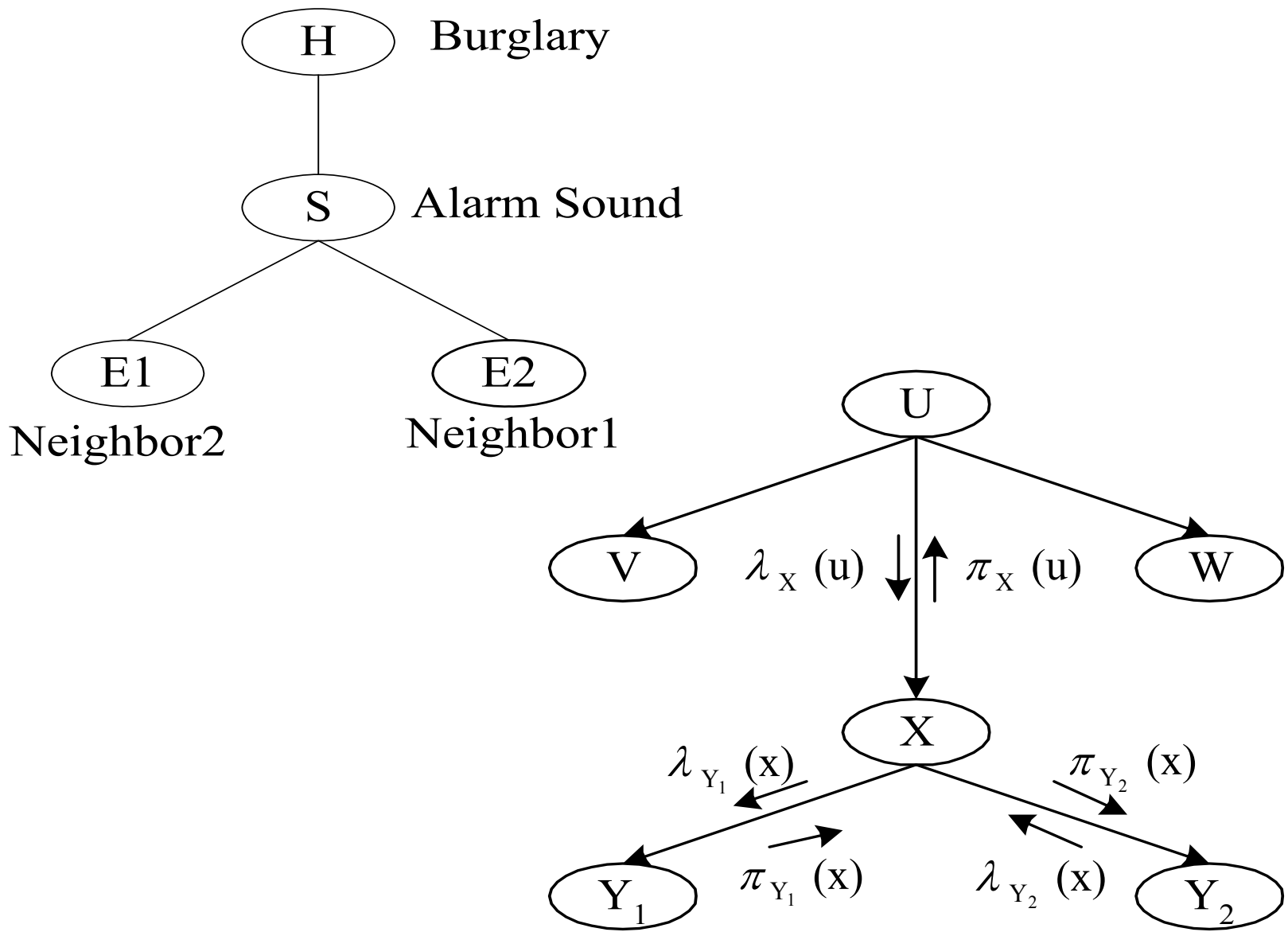


ECE 566

Bayesian Inference III





Initialization

1. Root Node: We set the $\pi(X)$ for a root node X to the prior probability for it : $\vec{P}(X)$
2. Anticipatory Nodes: If a node X is a childless node, we set $\lambda(X) = [1 \ 1 \ \dots \ 1]^T$.
3. Evidence Nodes: They don't receive π messages. They provide a λ vector. The λ node value is set to the likelihoods the evidence provides for each possible value of their parent node.

Belief Propagation and Belief Accumulation

Step 1: Belief Update

When node X is activated, it updates its own parameters, λ , π from λ supplied by its children and π supplied by its parent:

$$\lambda(X) = \prod_j \lambda_{Y_j}(X) \quad \text{term-wise product}$$

$$\pi(X) = M_{X|U} \pi_X(U)$$

$$\text{Bel}(X) = \alpha \lambda(X) \pi(X)$$

α : normalizing parameter to make sure $\sum_{X_i} \text{Bel}(X = X_i) = 1$

Step 2: Bottom-up Propagation

Using the λ messages received, node X computes a new λ message, $\lambda_X(U)$ which it sends to its parent U .

$$\begin{aligned}\lambda_X(U) &= \lambda(X) M_{X|U} \\ &= \sum_{X=X_i} \lambda(X) P(X = X_i | U)\end{aligned}$$

Step 3: Top down Propagation

Node X after updating itself, sends new π messages to each of its children:

$$\begin{aligned}\pi_{Y_j}(X) &= \alpha \pi(X) \prod_{k \neq j} \lambda_{Y_k}(X) \\ &= \frac{\text{Bel}(X)}{\lambda_{Y_j}(X)}\end{aligned}$$

Example

A murder has occurred and there are three suspects. The fingerprint lab provides information about fingerprints on the murder weapon.

Let X , and Y represent the following hypotheses:

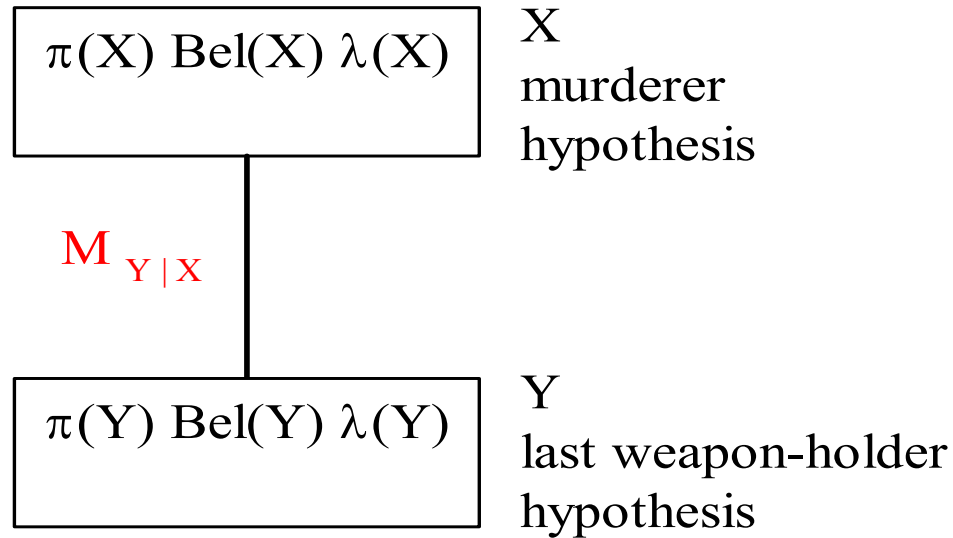
X : identity of the killer

Y : identity of the person who last held the weapon

Assuming that the conditional probability matrix $M_{Y|X}$ is as shown and the prior probability for each suspect is $\vec{P}(X) = [0.8 \ 0.1 \ 0.8]^T$

$$M_{Y|X} = \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.1 & 0.8 & 0.1 \\ 0.1 & 0.1 & 0.8 \end{bmatrix}$$

a) Show the belief network depicting this system and use the given information to initialize its parameters.



Initializing:

$$\pi(X) = P(X) = \begin{bmatrix} 0.8 \\ 0.1 \\ 0.1 \end{bmatrix}$$

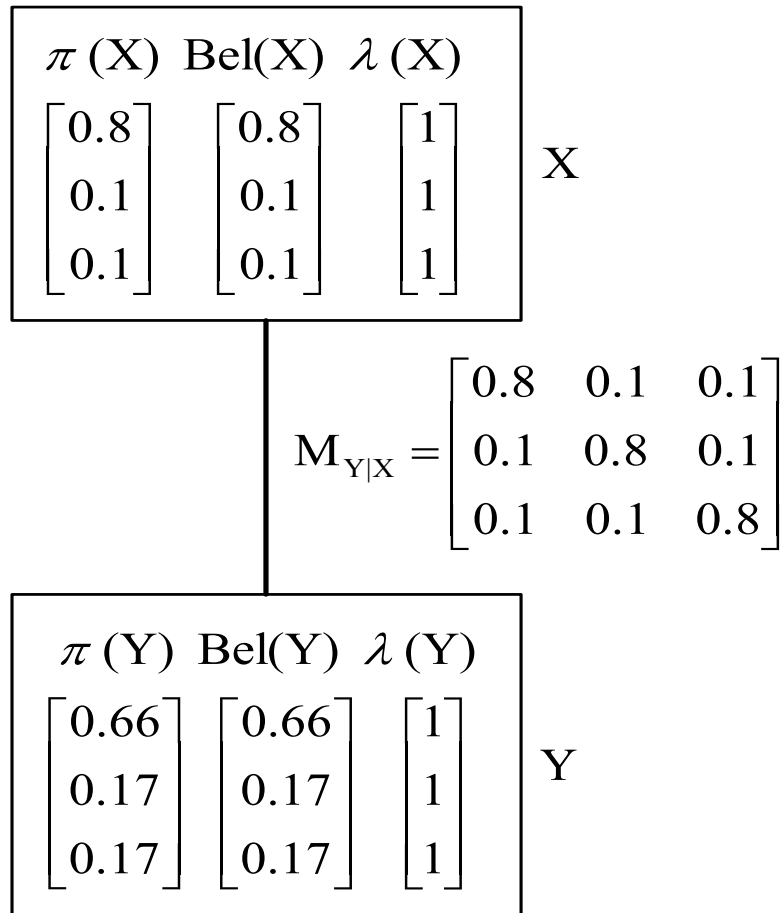
$$\lambda(Y) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad \text{anticipatory node}$$

$$\lambda(\mathbf{X}) = \frac{\text{Bel}(\mathbf{X})}{\pi(\mathbf{X})} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\begin{aligned} \pi(\mathbf{Y}) &= \pi_{\mathbf{X}}(\mathbf{Y}) \mathbf{M}_{\mathbf{Y}|\mathbf{X}} \\ &= \begin{bmatrix} 0.8 & 0.1 & 0.1 \end{bmatrix} \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.1 & 0.8 & 0.1 \\ 0.1 & 0.1 & 0.8 \end{bmatrix} \\ &= \begin{bmatrix} 0.66 \\ 0.17 \\ 0.17 \end{bmatrix} \end{aligned}$$

$$\text{Bel}(\mathbf{Y}) = \alpha \pi(\mathbf{Y}) \lambda(\mathbf{Y}) = \alpha \begin{bmatrix} 0.66 \\ 0.17 \\ 0.17 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.66 \\ 0.17 \\ 0.17 \end{bmatrix}$$

$(\alpha = 1)$



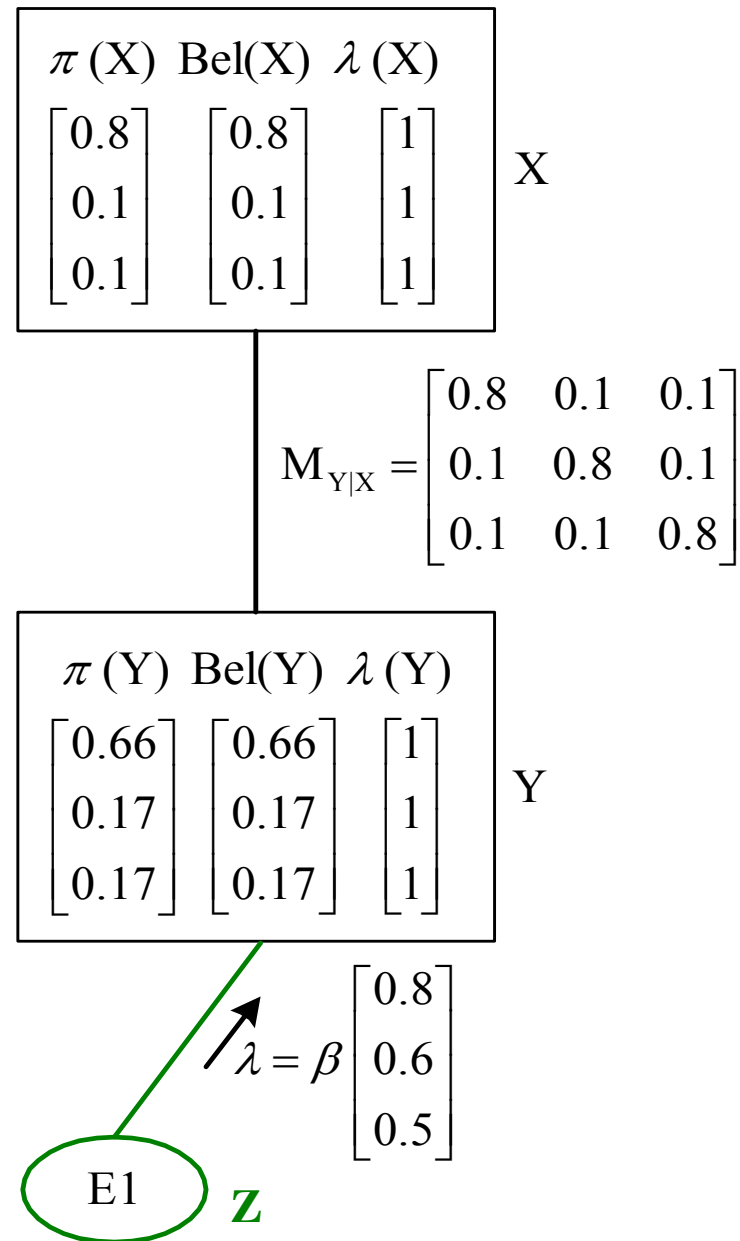
b) Assume the information (evidence) becomes available about fingerprints on the weapon. The evidence states the relative likelihood as $[0.8 \ 0.6 \ 0.5]^T$.

Propagate the evidence through the network and calculate the updated beliefs.

Let Z be the node representing the result from the fingerprint lab.

Z : an evidence node

Y receives a message from Z (a lab report) and updates itself.



$$\lambda_z(Y) = \beta \begin{bmatrix} 0.8 \\ 0.6 \\ 0.5 \end{bmatrix}$$

Updating node Y :

$$\lambda(Y) = \prod_i \lambda Y_i \quad \text{where } Y_i \text{ are children}$$

$$\lambda(Y) = \beta \begin{bmatrix} 0.8 \\ 0.6 \\ 0.5 \end{bmatrix}$$

$$\pi(Y) = \begin{bmatrix} 0.66 \\ 0.17 \\ 0.17 \end{bmatrix}$$

$$\Rightarrow \text{Bel}(Y) = \alpha \pi(Y) \lambda(Y)$$

$$\text{Bel}(Y) = \alpha \underbrace{\begin{bmatrix} 0.8 \\ 0.6 \\ 0.5 \end{bmatrix} \begin{bmatrix} 0.66 \\ 0.17 \\ 0.17 \end{bmatrix}}_{\text{product}} = \begin{bmatrix} 0.738 \\ 0.143 \\ 0.119 \end{bmatrix}$$

$$\begin{bmatrix} 0.528 \\ 0.102 \\ 0.085 \end{bmatrix} \Rightarrow \alpha = \frac{1}{0.528 + 0.102 + 0.085} = \frac{1}{0.715}$$

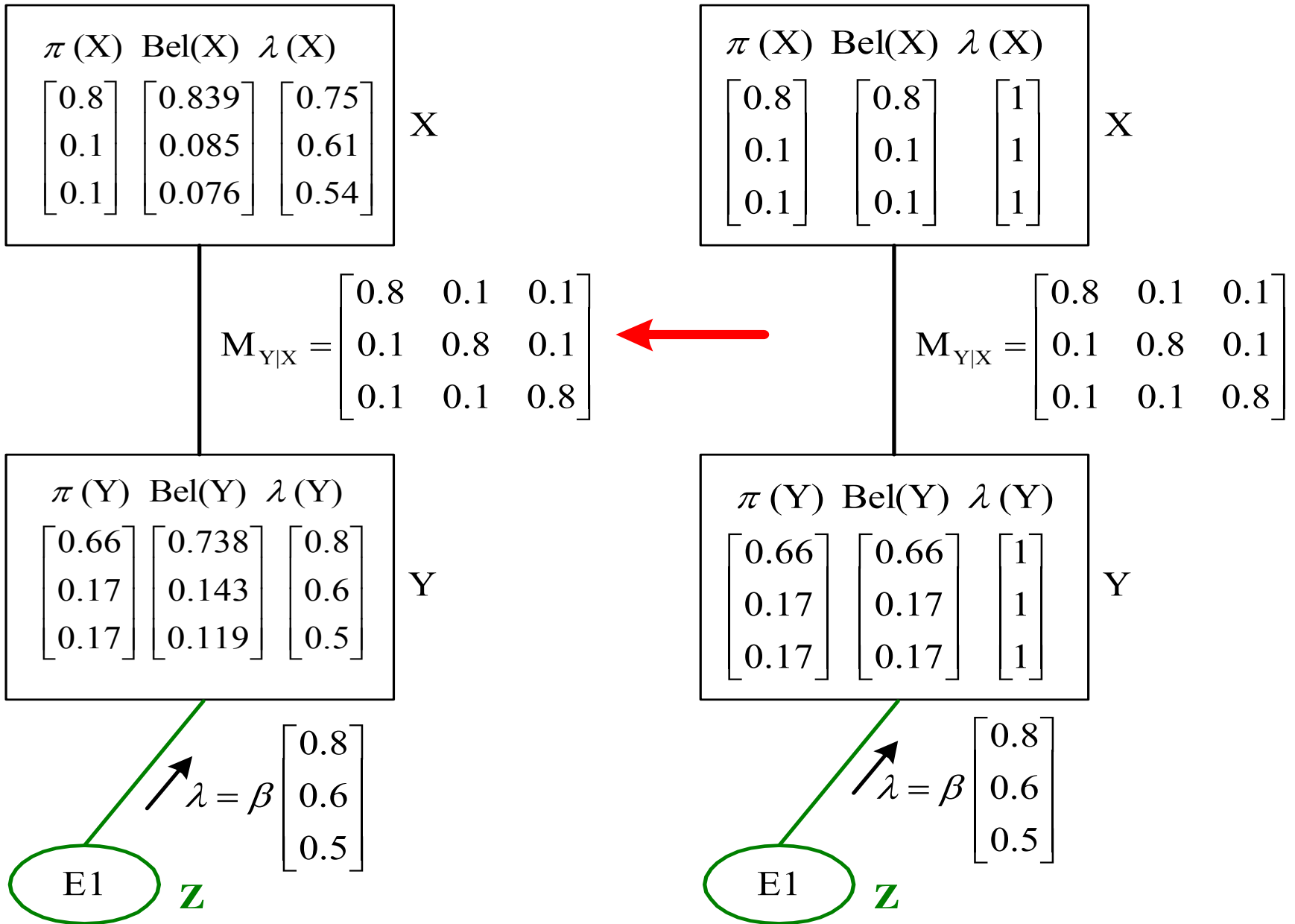
Bottom-up
propagation to X:

$$\lambda_Y(X) = M_{Y|X} \lambda(Y) = \beta \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.1 & 0.8 & 0.1 \\ 0.1 & 0.1 & 0.8 \end{bmatrix} \begin{bmatrix} 0.8 \\ 0.6 \\ 0.5 \end{bmatrix}$$
$$= \begin{bmatrix} 0.75 \\ 0.61 \\ 0.54 \end{bmatrix}$$

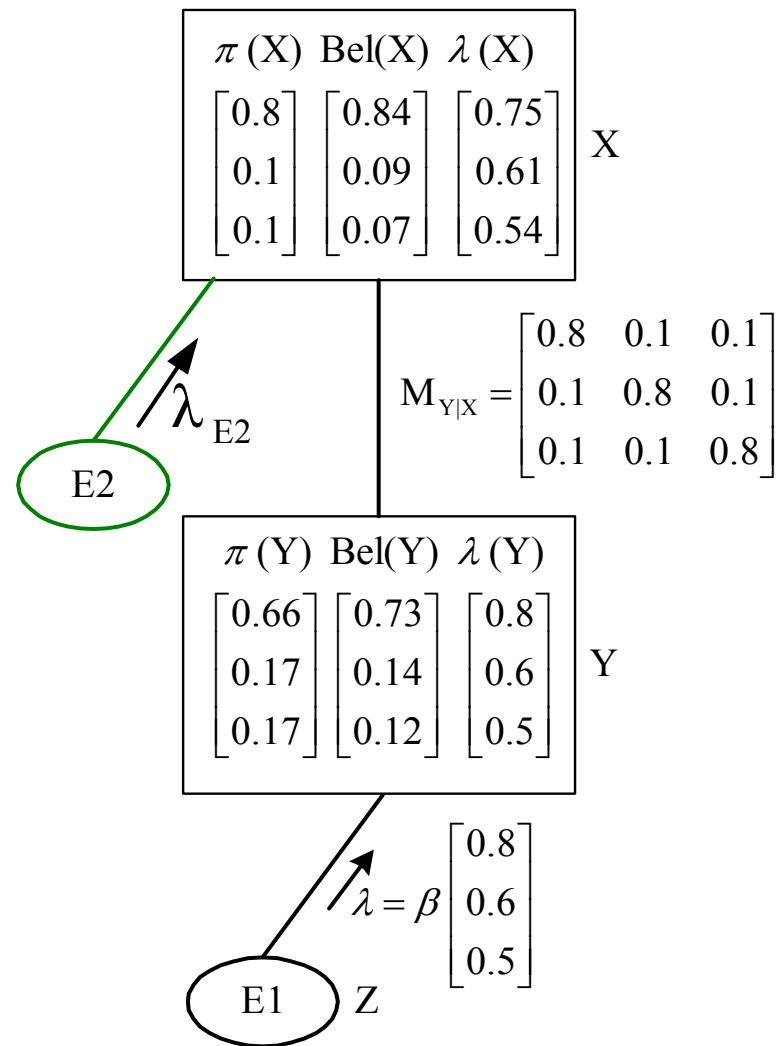
$$\lambda(X) = \prod_i \lambda_{Y_i}(X)$$
$$= \begin{bmatrix} 0.75 \\ 0.61 \\ 0.54 \end{bmatrix}$$

$$\pi(X) = \begin{bmatrix} 0.8 \\ 0.1 \\ 0.1 \end{bmatrix}$$

$$\text{Bel}(X) = \alpha \pi(X) \lambda(X)$$
$$= \alpha \begin{bmatrix} 0.8 \\ 0.1 \\ 0.1 \end{bmatrix} \begin{bmatrix} 0.75 \\ 0.61 \\ 0.54 \end{bmatrix} = \begin{bmatrix} 0.839 \\ 0.085 \\ 0.076 \end{bmatrix}$$



c) New information becomes available. Suspect 1 produces a strong alibi. Assume that this information provides the likelihoods $\beta_2[1 \ 10 \ 10]$ for identity of the killer. Integrate this information into the belief network and update the beliefs.



$$\lambda_{E2}(X) = \beta_2 \begin{bmatrix} 1 \\ 10 \\ 10 \end{bmatrix}$$

$$\lambda(X) = \lambda_{E2}(X) \lambda_Y(X) = \beta_2 \begin{bmatrix} 1 \\ 10 \\ 10 \end{bmatrix} \begin{bmatrix} 0.75 \\ 0.61 \\ 0.54 \end{bmatrix} = \beta_2 \begin{bmatrix} 0.75 \\ 6.1 \\ 5.4 \end{bmatrix}$$

$$\text{Bel}(X) = \alpha \lambda(X) \pi(X)$$

$$= \alpha \begin{bmatrix} 0.75 \\ 6.1 \\ 5.4 \end{bmatrix} \begin{bmatrix} 0.8 \\ 0.1 \\ 0.1 \end{bmatrix}$$

$$= \begin{bmatrix} 0.343 \\ 0.349 \\ 0.308 \end{bmatrix}$$

Top-down
propagation to Y:

$$\pi_Y(\mathbf{X}) = \alpha \pi(\mathbf{X}) \prod_{k \neq y} \lambda_k$$

$$\pi_Y(\mathbf{X}) = \alpha \begin{bmatrix} 0.8 \\ 0.1 \\ 0.1 \end{bmatrix} \begin{bmatrix} 1 \\ 10 \\ 10 \end{bmatrix} = \alpha \begin{bmatrix} 0.8 \\ 1 \\ 1 \end{bmatrix}$$

$$\pi(\mathbf{Y}) = \pi_{Y(\mathbf{X})} \mathbf{M}_{Y|\mathbf{X}}$$

$$= \alpha \begin{bmatrix} 0.8 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.1 & 0.8 & 0.1 \\ 0.1 & 0.1 & 0.8 \end{bmatrix}$$

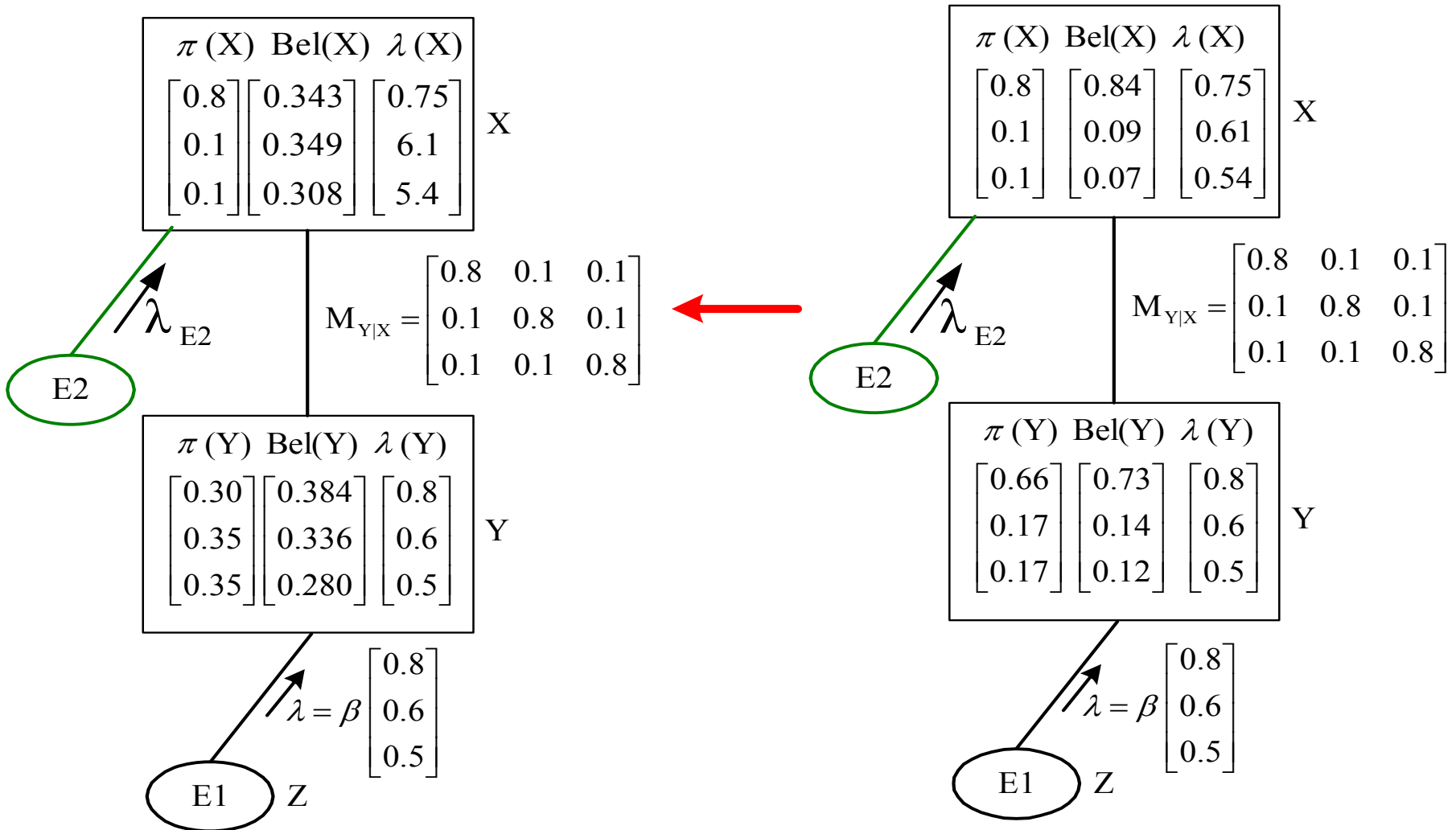
$$= \begin{bmatrix} 0.30 \\ 0.35 \\ 0.35 \end{bmatrix}$$

$$\lambda(\mathbf{Y}) = \begin{bmatrix} 0.8 \\ 0.6 \\ 0.5 \end{bmatrix}$$

$$\text{Bel}(\mathbf{Y}) = \alpha \pi(\mathbf{Y}) \lambda(\mathbf{Y})$$

$$= \alpha \begin{bmatrix} 0.30 \\ 0.35 \\ 0.35 \end{bmatrix} \begin{bmatrix} 0.8 \\ 0.6 \\ 0.5 \end{bmatrix}$$

$$= \begin{bmatrix} 0.384 \\ 0.336 \\ 0.280 \end{bmatrix}$$



The new information reduces the likelihood that the murderer is suspect 1 from 0.84 to 0.343.