

Random variables in the graph are defined by:

$$\begin{split} P(R) &= [P(R = '\text{low'}) \quad P(R = '\text{high'})]^{\text{T}} = [30\% \quad 70\%]^{\text{T}} \\ P(S) &= [P(S = '\text{low'}) \quad P(S = '\text{high'})]^{\text{T}} = [70\% \quad 30\%]^{\text{T}} \\ P(D|S,R = '\text{low'}) &= \begin{bmatrix} P(D = '\text{low'}|S = '\text{low'}, R = '\text{low'}) & P(D = '\text{low'}|S = '\text{high'}, R = '\text{low'}) \\ P(D = '\text{high'}|S = '\text{low'}, R = '\text{low'}) & P(D = '\text{high'}|S = '\text{high'}, R = '\text{low'}) \end{bmatrix} \\ &= \begin{bmatrix} 85\% \quad 50\% \\ 15\% \quad 50\% \end{bmatrix} \end{split}$$

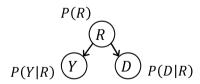
$$P(D|S,R = '\text{high'}) = \begin{bmatrix} P(D = '\text{low'}|S = '\text{low'},R = '\text{high'}) & P(D = '\text{low'}|S = '\text{high'},R = '\text{high'}) \\ P(D = '\text{high'}|S = '\text{low'},R = '\text{high'}) & P(D = '\text{high'}|S = '\text{high'},R = '\text{high'}) \end{bmatrix}$$

$$= \begin{bmatrix} 100\% & 90\% \\ 0\% & 10\% \end{bmatrix}$$

$$P(Y|R) = \begin{bmatrix} P(Y = '\text{high'}|R = '\text{high'}) & P(Y = '\text{high'}|R = '\text{low'}) \\ P(Y = '\text{low'}|R = '\text{high'}) & P(Y = '\text{low'}|R = '\text{low'}) \end{bmatrix} = \begin{bmatrix} 95\% & 5\% \\ 5\% & 95\% \end{bmatrix}$$

$$P(Y|R) = \begin{bmatrix} P(Y = '\text{high'}|R = '\text{high'}) & P(Y = '\text{high'}|R = '\text{low'}) \\ P(Y = '\text{low'}|R = '\text{high'}) & P(Y = '\text{low'}|R = '\text{low'}) \end{bmatrix} = \begin{bmatrix} 95\% & 5\% \\ 5\% & 95\% \end{bmatrix}$$

We can eliminate variable S, finding conditional probability P(D|R):



to do so, we compute $P(D|R = 'low') = [P(D = 'low'|R = 'low') \quad P(D = 'high'|R = 'low')]^T =$ $\sum_{S} P(D|S, R = 'low')P(S)$

by multiplying matrix P(D|S, R = 'low') and vector P(S), we get

$$P(D|R = 'low') = [74.5\% 25.5\%]^{T}$$

and

$$P(D|R = '\text{high'}) = [P(D = '\text{low'}|R = '\text{high'}) \quad P(D = '\text{high'}|R = '\text{high'})]^{T}$$

$$= \sum_{S} P(D|S, R = '\text{high'})P(S)$$

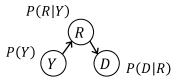
by multiplying matrix P(D|S, R = 'high') and vector P(S), we get

$$P(D|R = 'high') = [97\% 3\%]^{T}$$

By arranging the two vectors in a single matrix, we obtain:

So
$$P(D|R) = \begin{bmatrix} P(D = 'low'|R = 'low') & P(D = 'low'|R = 'high') \\ P(D = 'high'|R = 'low') & P(D = 'high'|R = 'high') \end{bmatrix} = \begin{bmatrix} 74.5\% & 97\% \\ 25.5\% & 3\% \end{bmatrix}$$

We invert the link between Y and R:



Marginal of *Y* is:

$$P(Y) = [P(Y = 'low') \quad P(Y = 'high')]^{T} = \sum_{R} P(Y|R)P(R) = [32\% \quad 68\%]^{T}$$

by multiplying matrix P(Y|R) and vector P(R).

Conditional of R given Y is

$$P(R|Y) = \begin{bmatrix} P(R = 'low'|Y = 'low') & P(R = 'low'|Y = 'high') \\ P(R = 'high'|Y = 'low') & P(R = 'high'|Y = 'high') \end{bmatrix} = \frac{P(Y|R)P(R)}{P(Y)} = \begin{bmatrix} 89.1\% & 2.2\% \\ 10.9\% & 97.8\% \end{bmatrix}$$

by multiplying matrix P(Y|R) element by element by vector P(R) and dividing by vector P(Y).

Now we marginalize variable R:

$$P(Y)$$
 Y D $P(D|Y)$

Probability P(D|Y) is given by:

$$P(D|Y) = \begin{bmatrix} P(D = 'low'|Y = 'low') & P(D = 'low'|Y = 'high') \\ P(D = 'high'|Y = 'low') & P(D = 'high'|Y = 'high') \end{bmatrix} = \sum_{R} P(D|R)P(R|Y)$$
$$= \begin{bmatrix} 77.0\% & 96.5\% \\ 23.0\% & 3.5\% \end{bmatrix}$$

where we get the matrix by matrix multiplication.

Marginal distribution of damage is:

$$P(D) = \sum_{Y} P(D|Y)P(Y) = [P(D = 'low') \quad P(D = 'high')]^{T} = [90.2\% \quad 9.8\%]^{T}$$

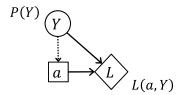
where we get the vector by matrix-vector multiplication.

So probability of failure is P(D = 'high') = 9.8%, it goes up to P(D = 'high'|Y = 'low') = 23.0%, or down to P(D = 'high'|Y = 'high') = 3.5%, depending to the observed value of Y.

The loss matrix is

$$L(a,D) = \begin{bmatrix} L(a = \text{'do nothing'}, D = \text{'low'}) & L(a = \text{'do nothing'}, D = \text{'high'}) \\ L(a = \text{'repair'}, D = \text{'low'}) & L(a = \text{'repair'}, D = \text{'high'}) \end{bmatrix} = \begin{bmatrix} 0 & 100 \\ 20 & 25 \end{bmatrix}$$

The corresponding loss matrix depending by Y is



We compute the matrix as:

$$L(a,Y) = \begin{bmatrix} L(a = \text{'do nothing'}, Y = \text{'low'}) & L(a = \text{'do nothing'}, Y = \text{'high'}) \\ L(a = \text{'repair'}, Y = \text{'low'}) & L(a = \text{'repair'}, Y = \text{'high'}) \end{bmatrix} = \sum_{R} L(a,D)P(D|Y)$$

$$= \begin{bmatrix} 23.0 & 3.5 \\ 21.1 & 20.1 \end{bmatrix}$$

The expected loss relating to action a is:

$$EL(a) = [L(a = \text{'do nothing'}) \quad L(a = \text{'repair'})]^{T} = \sum_{Y} L(a, Y)P(Y) = [9.75 \quad 20.49]^{T}$$

The optimal action is that minimizing the expected loss:

$$a^* = \underset{a}{\operatorname{argmin}} EL(a) = \text{`do nothing'}$$

and the corresponding loss is

$$EL^* = \min_a EL(a) = 9.75$$

If Y = 'low', the optimal action becomes

$$a^*(Y = \text{'low'}) = \underset{a}{\operatorname{argmin}} L(a, Y = \text{'low'}) = \text{'repair'}$$

and the corresponding loss is

$$L^*(Y = \text{'low'}) = \min_{a} L(a, Y = \text{'low'}) = 21.1$$

If Y = 'high', the optimal action is

$$a^*(Y = \text{'high'}) = \underset{a}{\operatorname{argmin}} L(a, Y = \text{'high'}) = \text{'do nothing'}$$

and the corresponding loss is

$$L^*(Y = \text{'high'}) = \min_{a} L(a, Y = \text{'high'}) = 3.5$$

The expected loss observing Y is:

$$EL_{\text{obs}}^* = \sum_{Y} L^*(Y)P(Y) = 21.1 \cdot 32\% + 3.5 \cdot 68\% = 9.14$$

and the Value of information is

$$VoI = EL^* - EL_{\rm obs}^* = 0.60$$