## The max-min-hill-climbing algorithm

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Letter

#### Definition (Conditional Independence)

Two variables X and Y are conditionally independent given  $\mathbf{Z}$  w.r.t a probability distribution P, denoted as  $Ind_p(X; Y|\mathbf{Z})$ , if  $\forall x, y, \mathbf{z}$ , where  $P(\mathbf{Z} = \mathbf{z}) > 0$ ,

$$P(X, Y|\mathbf{Z}) = P(X|\mathbf{Z}) * P(Y|\mathbf{Z}), \tag{1}$$

where  $P(X, Y|\mathbf{Z}) = P(X \cap Y|\mathbf{Z})$ .

It is equivalent

$$Ind(X; Y|\mathbf{Z}) \Longleftrightarrow (Assoc(X; Y|\mathbf{Z}) = 0), \tag{2}$$

where  $Assoc(X; Y|\mathbf{Z})$  is the strength association (dependency) of X and Y given  $\mathbf{Z}$ .

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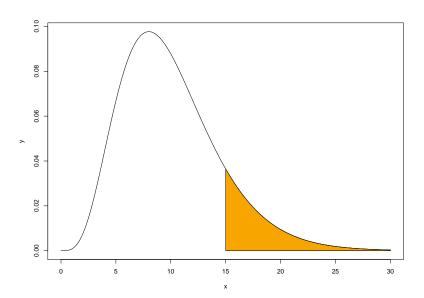
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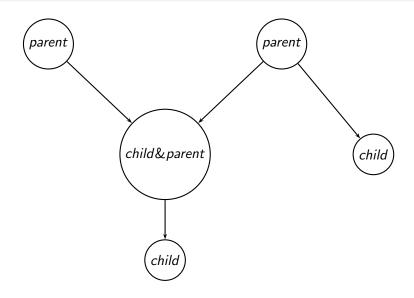
#### **Definition**

We calculate the  $G^2$  value under the nullhypothesis of the conditional independence of  $Ind_P(X_i, X_j | \mathbf{X}_k)$  holding. Then, the  $G^2$  value is defined as:

$$G^{2} := 2 * \sum_{a,b,c} S_{ijk}^{abc} * In \left( \frac{S_{ijk}^{abc} * S_{k}^{c}}{S_{ik}^{ac} * S_{jk}^{bc}} \right)$$
(3)



# Bayesian Networks



### Collider



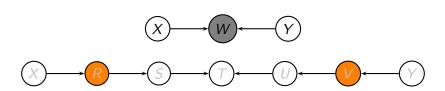
### Blocked paths

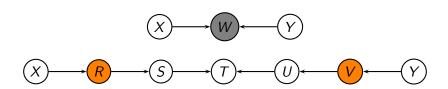
#### Define four sets

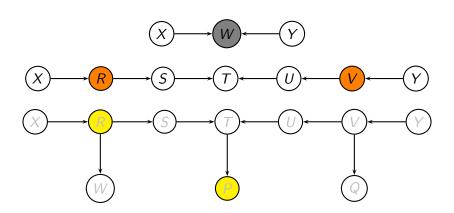
Let  $A_1$ ,  $A_2$  and  $A_3$  denote sets with:

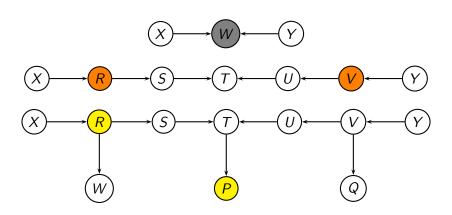
$$A_1 = \{W\}$$
  
 $A_2 = \{R, V\}$   
 $A_3 = \{R, P\}$ 

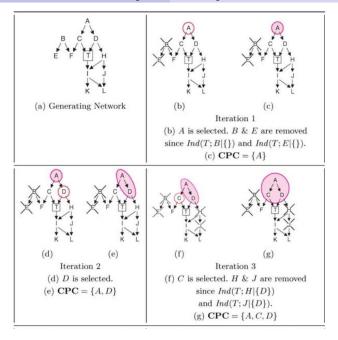


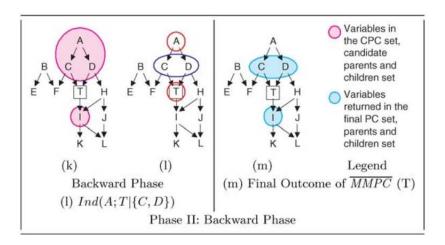












# Thanks for your attention!