

Elisabetta Fersini

## ***Esercitazione***

DISCo

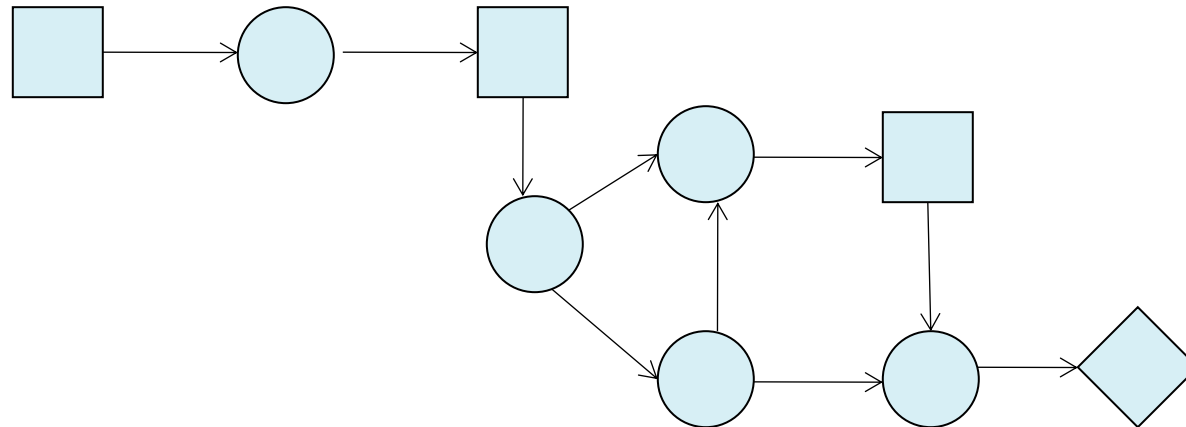
Università degli Studi di Milano-

Bicocca

Viale Sarca, 336

20126 Milano

[elisabetta.fersini@unimib.it](mailto:elisabetta.fersini@unimib.it)

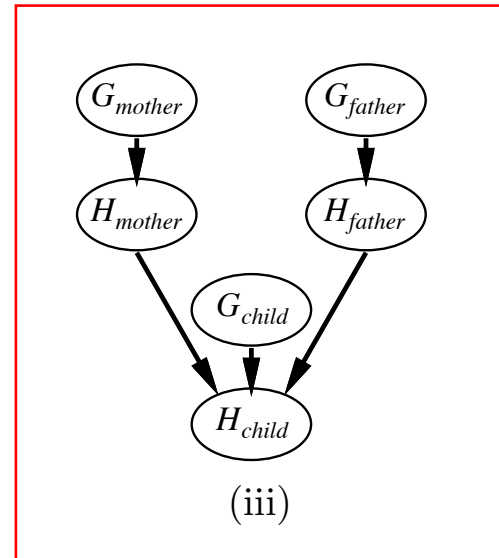
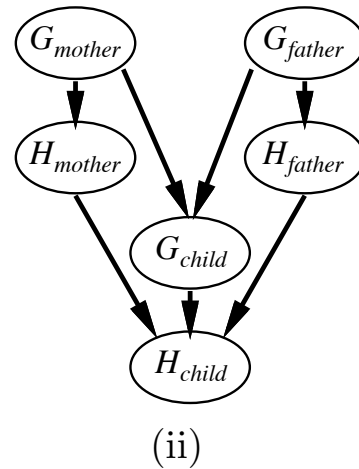
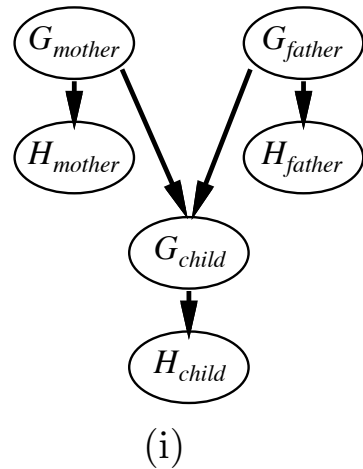


# Esercizio 1

- Sia  $H_x$  una variabile casuale che denota la manualità di un individuo  $x$ , la quale può assumere valore  $L=Left$  e  $R=Right$ . Un'ipotesi comune è che l'essere destrorsi ( $R$ ) o mancini ( $L$ ) sia ereditato da un semplice meccanismo: c'è un gene  $G_x$ , che assume valori  $L$  or  $R$ , e con probabilità  $s$  l'individuo assume la stessa manualità del gene. Inoltre, il gene stesso ha uguale probabilità di essere ereditato da un genitore, ma con probabilità  $m$  può ereditare una manualità opposta a quella dei genitori.

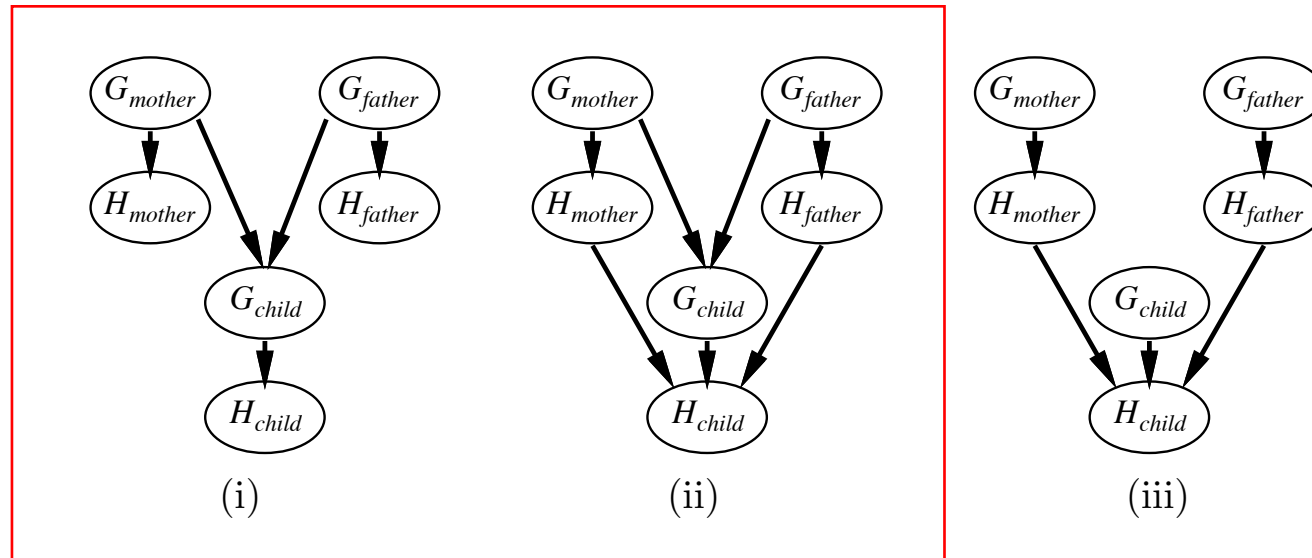
# Esercizio 1

1. Quali delle tre reti asserisce che  $P(G_{\text{father}}, G_{\text{mother}}, G_{\text{child}}) = P(G_{\text{father}})P(G_{\text{mother}})P(G_{\text{child}})$ ?



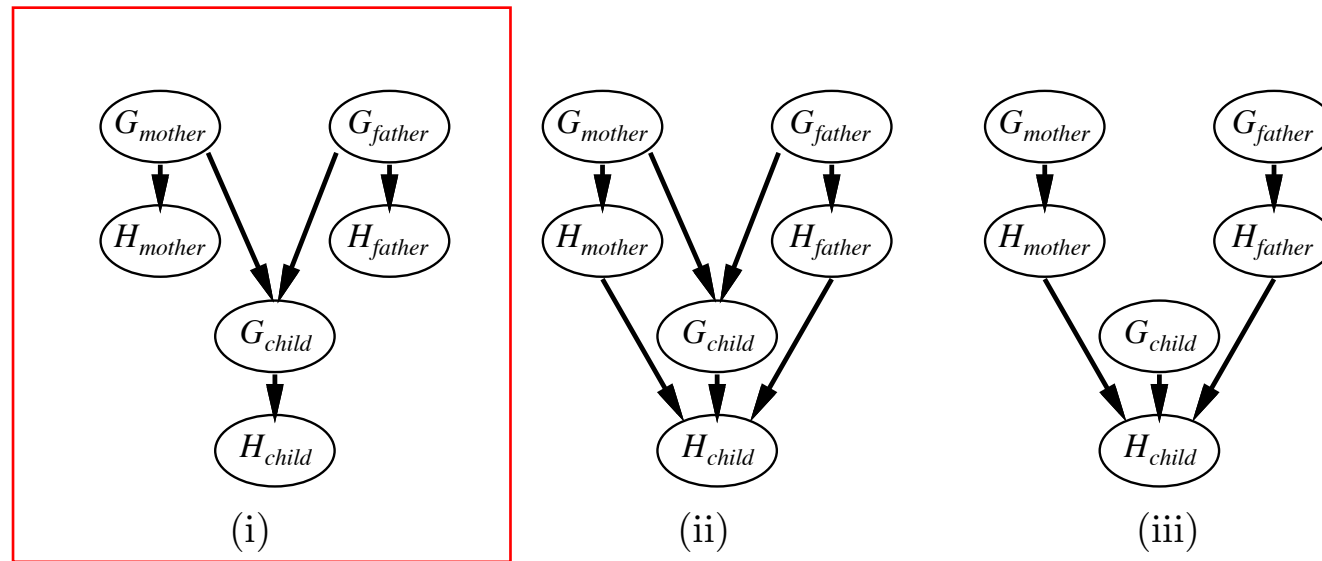
# Esercizio 1

2. Quali delle tre reti bayesiane afferma le condizioni di indipendenza coerenti con l'ipotesi descritta nel testo del problema?



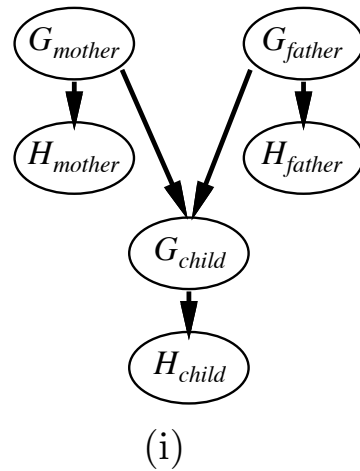
# Esercizio 1

3. Quale delle tre reti descrive al meglio l'ipotesi descritta nel problema?



# Esercizio 1

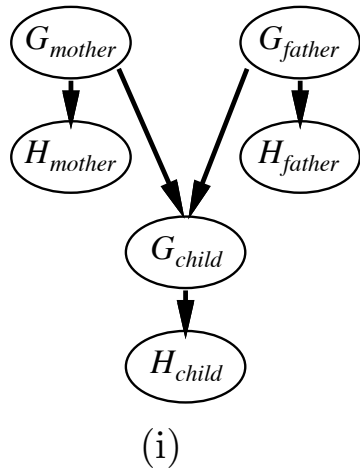
4. Scrivere una CPT per  $G_{child}$  relative alla rete (i) o (ii), coerente con le specifiche del problema.



$G_{mother}$	$G_{father}$	$P(G_{child}   G_{mother}, G_{father})$	
		L	R
L	L		
L	R		
R	L		
R	R		

# Esercizio 1

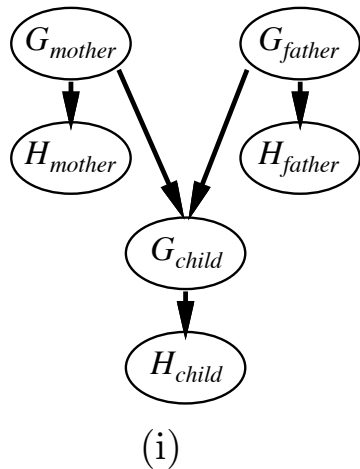
5. Supponiamo che  $P(G_{\text{father}} = L) = P(G_{\text{mother}} = L) = x$ . Si derivi per la rete (i) or (ii), l'espressione per  $P(G_{\text{child}} = L)$  in termini di  $m$  ed  $x$ , condizionando solo rispetto ai nodi genitori.



$$\begin{aligned} P(G_{\text{child}} = L) &= \sum_{G_{\text{father}}} \sum_{G_{\text{mother}}} P(G_{\text{father}}) * P(G_{\text{mother}}) * P(G_{\text{child}} | G_{\text{father}}, G_{\text{mother}}) \\ &= P(G_{\text{father}} = L) * P(G_{\text{mother}} = L) * P(G_{\text{child}} = L | G_{\text{father}} = L, G_{\text{mother}} = L) + \\ &\quad + P(G_{\text{father}} = R) * P(G_{\text{mother}} = R) * P(G_{\text{child}} = L | G_{\text{father}} = R, G_{\text{mother}} = R) + \\ &\quad + P(G_{\text{father}} = L) * P(G_{\text{mother}} = R) * P(G_{\text{child}} = L | G_{\text{father}} = L, G_{\text{mother}} = R) + \\ &\quad + P(G_{\text{father}} = R) * P(G_{\text{mother}} = L) * P(G_{\text{child}} = L | G_{\text{father}} = R, G_{\text{mother}} = L) \end{aligned}$$

# Esercizio 1

5. Supponiamo che  $P(G_{\text{father}} = L) = P(G_{\text{mother}} = L) = x$ . Si derivi per la rete (i) or (ii), l'espressione per  $P(G_{\text{child}} = L)$  in termini di  $m$  ed  $x$ , condizionando solo rispetto ai nodi genitori.

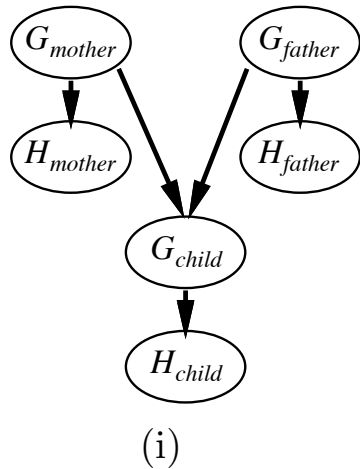


$$\begin{aligned}
 & \boxed{x^2} \quad \boxed{(1-m)} \\
 = & \boxed{P(G_{\text{father}} = L) * P(G_{\text{mother}} = L)} * \boxed{P(G_{\text{child}} = L | G_{\text{father}} = L, G_{\text{mother}} = L)} + \\
 & \boxed{(1-x^2)} \quad \boxed{m} \\
 + & \boxed{P(G_{\text{father}} = R) * P(G_{\text{mother}} = R)} * \boxed{P(G_{\text{child}} = L | G_{\text{father}} = R, G_{\text{mother}} = R)} + \\
 & \boxed{(x) * (1-x)} \quad \boxed{0.5} \\
 + & \boxed{P(G_{\text{father}} = L) * P(G_{\text{mother}} = R)} * \boxed{P(G_{\text{child}} = L | G_{\text{father}} = L, G_{\text{mother}} = R)} + \\
 & \boxed{(1-x) * (x)} \quad \boxed{0.5} \\
 + & \boxed{P(G_{\text{father}} = R) * P(G_{\text{mother}} = L)} * \boxed{P(G_{\text{child}} = L | G_{\text{father}} = R, G_{\text{mother}} = L)}
 \end{aligned}$$



# Esercizio 1

5. Supponiamo che  $P(G_{\text{father}} = L) = P(G_{\text{mother}} = L) = x$ . Si derivi per la rete (i) or (ii), l'espressione per  $P(G_{\text{child}} = L)$  in termini di  $m$  ed  $x$ , condizionando solo rispetto ai nodi genitori.



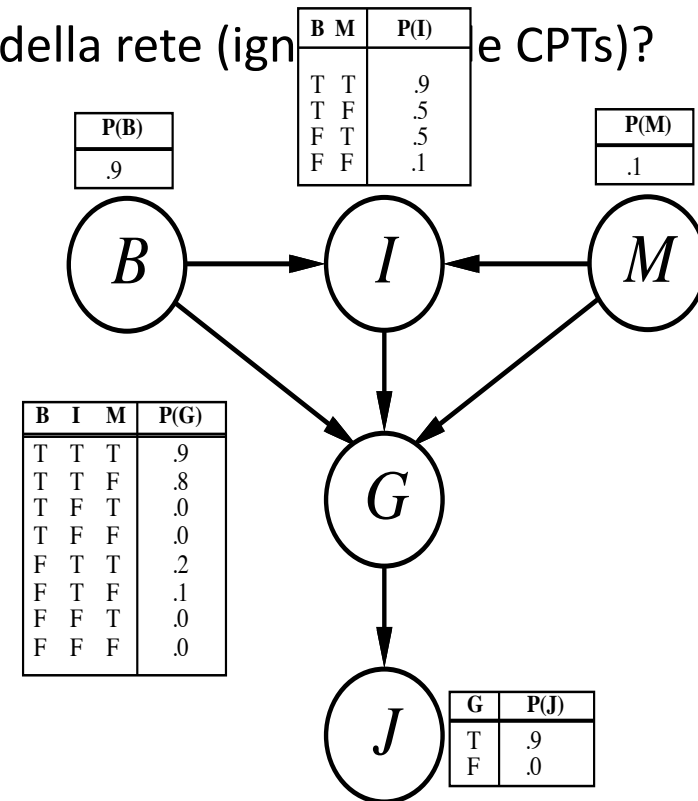
$$\begin{aligned} P(G_{\text{child}} = L) &= \sum_{G_{\text{father}}} \sum_{G_{\text{mother}}} P(G_{\text{father}}) * P(G_{\text{mother}}) * P(G_{\text{child}} | G_{\text{father}}, G_{\text{mother}}) \\ &= (x^2 - mx^2) + (m + mx^2 - 2mx) + 0.5x - 0.5x^2 + 0.5x - 0.5x^2 = \\ &= x^2 - mx^2 + m + mx^2 - 2mx + 0.5x - 0.5x^2 + 0.5x - 0.5x^2 = \\ &= m + x - 2mx \end{aligned}$$

# Esercizio 2

- Si consideri la seguente Rete Bayesiana.

1. Quali delle seguenti affermazioni è asserita dalla struttura della rete (ignorando le CPTs)?

- $P(B, I, M) = P(B)P(I)P(M)$  **falso**
- $P(J|G) = P(J|G, I)$  **vero**
- $P(M|G, B, I) = P(M|G, B, I, J)$  **vero**

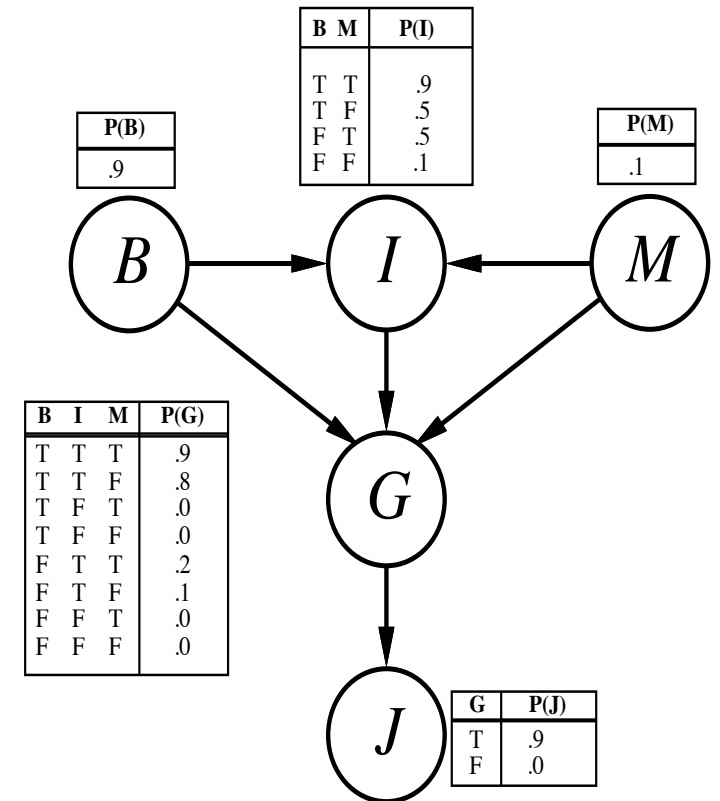


# Esercizio 2

2. Calcolare il valore di  $P(b, i, \neg m, g, j)$ .

$$P(b, i, \neg m, g, j) = P(b) * P(\neg m) * P(i|b, \neg m) * P(g|b, i, \neg m) * P(j|g) =$$

$$= 0.9 * 0.9 * 0.5 * 0.8 * 0.9 = 0.291$$



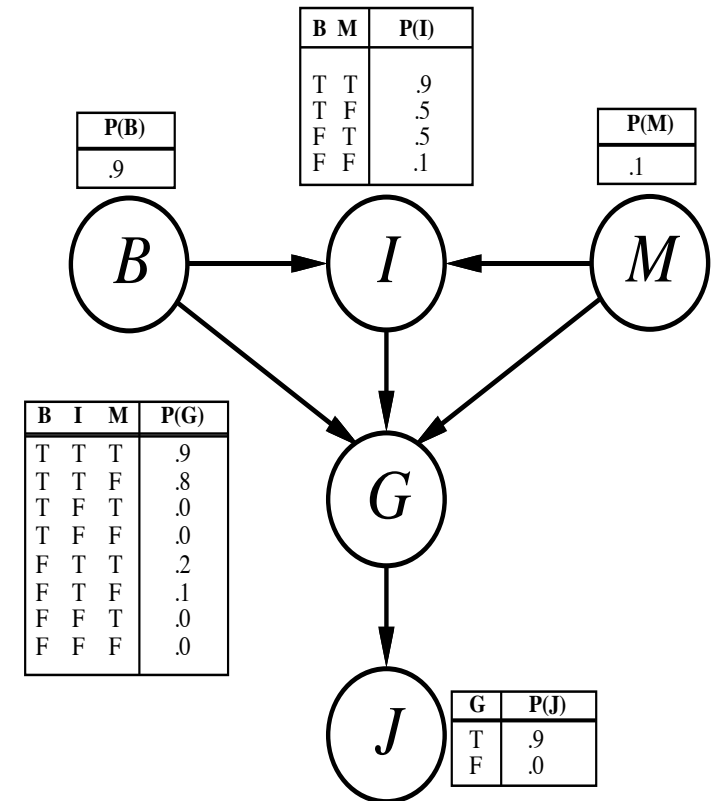
# Esercizio 2

3. Calcolare  $P(J|b,i,m)$

$$P(J|b,i,m) = \alpha \sum_G P(J, b, i, m, G)$$

$\uparrow$   $\uparrow$

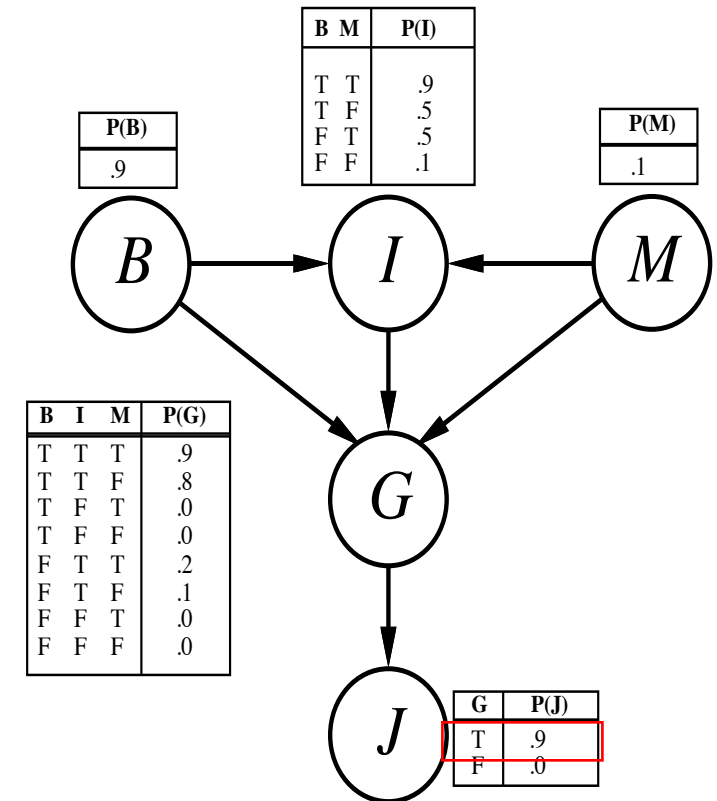
$P(j|b,i,m)$   $P(\neg j|b,i,m)$



# Esercizio 2

3. Calcolare  $P(J|b,i,m)$

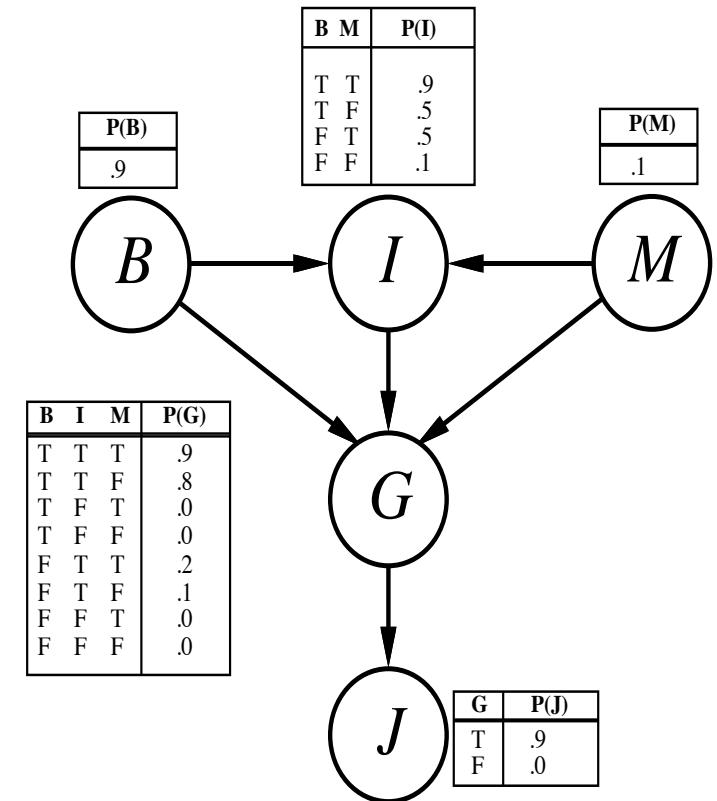
$$\begin{aligned}
 P(j|b,i,m) &= \alpha \sum_G P(j,b,i,m,G) = \\
 &= \alpha \sum_G P(j|G) * P(G|b,i,m) * P(b) * P(m) * P(i|b,m) \\
 &= \alpha * [P(j|g) * P(g|b,i,m) * P(b) * P(m) * P(i|b,m) + \\
 &\quad + P(j|\neg g) * P(\neg g|b,i,m) * P(b) * P(m) * P(i|b,m)] \\
 &= \alpha * [0.9 * 0.9 * 0.9 * 0.1 * 0.9 + 0 * 0.1 * 0.0 * 0.1 * 0.9] \\
 &= \alpha * [0.06561]
 \end{aligned}$$



# Esercizio 2

3. Calcolare  $P(J|b,i,m)$

$$\begin{aligned}
 P(\neg j|b, i, m) &= \alpha \sum_G P(\neg j, b, i, m, G) = \\
 &= \alpha \sum_G P(\neg j|G) * P(G|b, i, m) * P(b) * P(m) * P(i|b, m) \\
 &= \alpha * [P(\neg j|g) * P(g|b, i, m) * P(b) * P(m) * P(i|b, m) + \\
 &\quad + P(\neg j|\neg g) * P(\neg g|b, i, m) * P(b) * P(m) * P(i|b, m)] \\
 &= \alpha * [0.1 * 0.9 * 0.9 * 0.1 * 0.9 + 1 * 0.1 * 0.0 * 0.1 * 0.9] \\
 &= \alpha * [0.01539]
 \end{aligned}$$



# Esercizio 2

3. Calcolare  $P(J|b,i,m)$

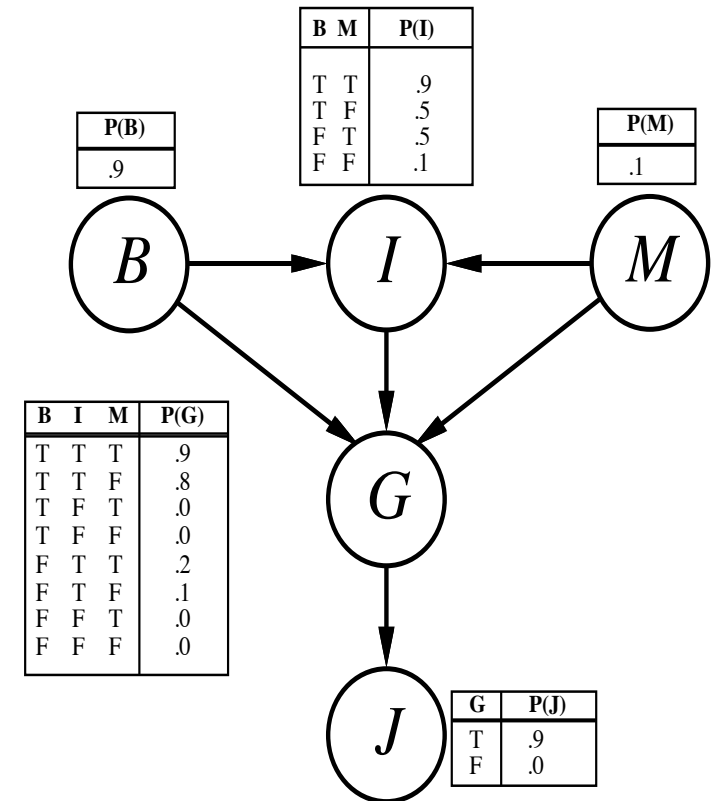
$$P(j|b, i, m) = \alpha * [0.06561]$$

$$P(\neg j|b, i, m) = \alpha * [0.01539]$$

$$P(J|b, i, m) = \alpha * \langle 0.06561; 0.01539 \rangle$$

$$\alpha = \frac{1}{0.06561 + 0.01539} = 12,3456790123$$

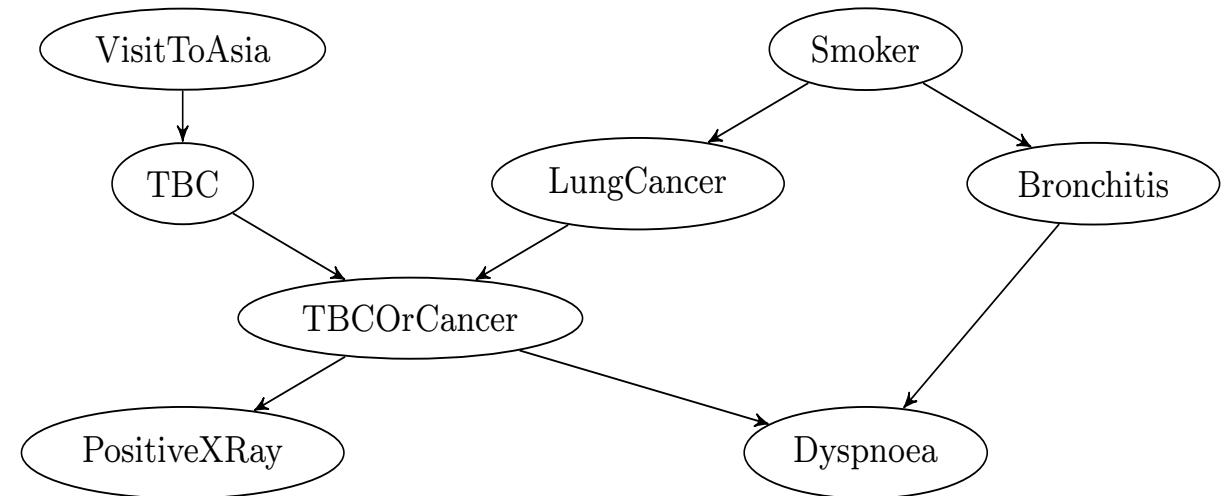
$$P(J|b, i, m) = \langle 0.81; 0.19 \rangle$$



# Esercizio 3

- Si consideri la seguente Rete Bayesiana. Determine which of the following conditional independence statements follow from the structure of the Bayesian network:

- (i)  $\text{Ind}(\text{TBC}, \text{VisitToAsia})$
- (ii)  $\text{Ind}(\text{VisitToAsia}, \text{Smoker})$
- (iii)  $\text{Ind}(\text{VisitToAsia}, \text{PositiveXRay} | \text{TBCOrCancer})$
- (iv)  $\text{Ind}(\text{VisitToAsia}, \text{Dyspnoea} | \text{TBCOrCancer})$
- (v)  $\text{Ind}(\text{TBC}, \text{Smoker} | \text{PositiveXRay})$





# Esercizio 3

[True/False]: Supponiamo che le variabili  $X_1, X_2, \dots, X_k$  non abbiano genitore in una determinata Rete Bayesiana che contiene  $n$  variabili in tutto, dove  $n > k$ . La Rete Bayesiana asserisce che  $P(X_1, X_2, \dots, X_k) = P(X_1)P(X_2) \cdots P(X_k)$ .

vero

# Esercizio 4

- Calcolare  $P(\text{Dyspnoea} | \text{Smoker}, \neg \text{TBC})$ . The entries necessarie delle CPT sono le seguenti:

- $P(\text{LungCancer} | \text{Smoker}) = 0.1$
- $P(\text{LungCancer} | \neg \text{Smoker}) = 0.01$
- $P(\text{Bronchitis} | \text{Smoker}) = 0.2$
- $P(\text{Bronchitis} | \neg \text{Smoker}) = 0.1$
- $P(\text{TBCOrCancer} | \text{TBC}, \text{LungCancer}) = 1$
- $P(\text{TBCOrCancer} | \text{TBC}, \neg \text{LungCancer}) = 1$
- $P(\text{TBCOrCancer} | \neg \text{TBC}, \text{LungCancer}) = 1$
- $P(\text{TBCOrCancer} | \neg \text{TBC}, \neg \text{LungCancer}) = 0$
- $P(\text{Dyspnoea} | \text{TBCOrCancer}, \text{Bronchitis}) = 0.9$
- $P(\text{Dyspnoea} | \text{TBCOrCancer}, \neg \text{Bronchitis}) = 0.7$
- $P(\text{Dyspnoea} | \neg \text{TBCOrCancer}, \text{Bronchitis}) = 0.6$
- $P(\text{Dyspnoea} | \neg \text{TBCOrCancer}, \neg \text{Bronchitis}) = 0.05$

