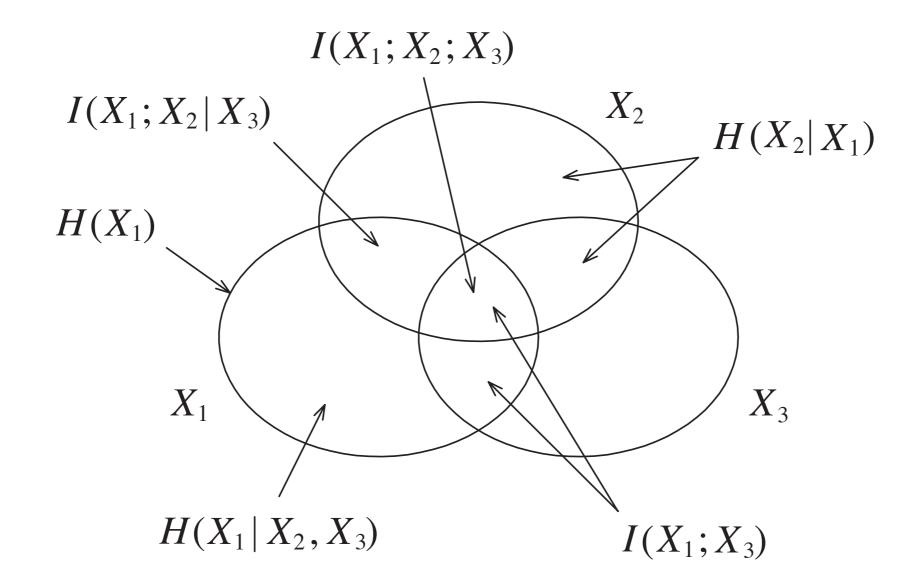
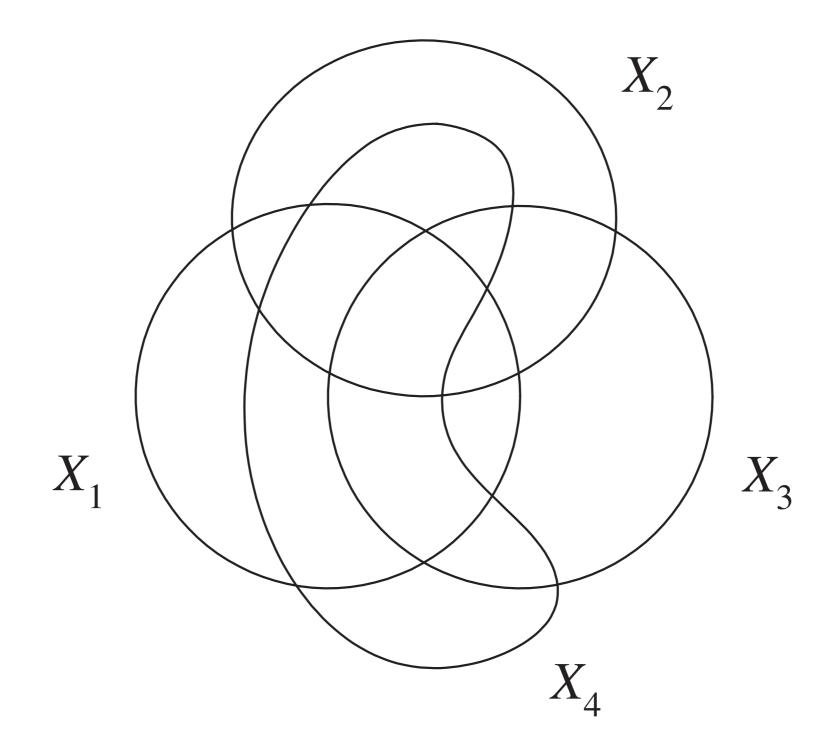
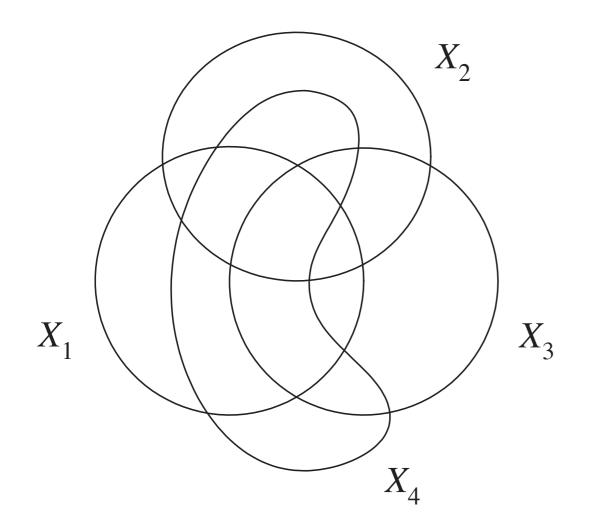
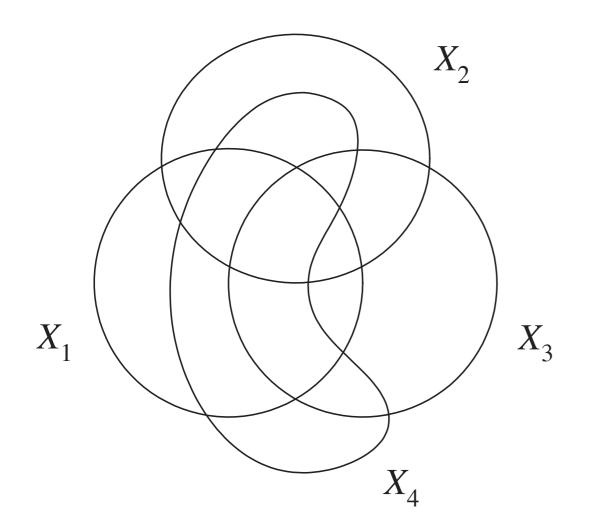


# 3.5 Information Diagrams

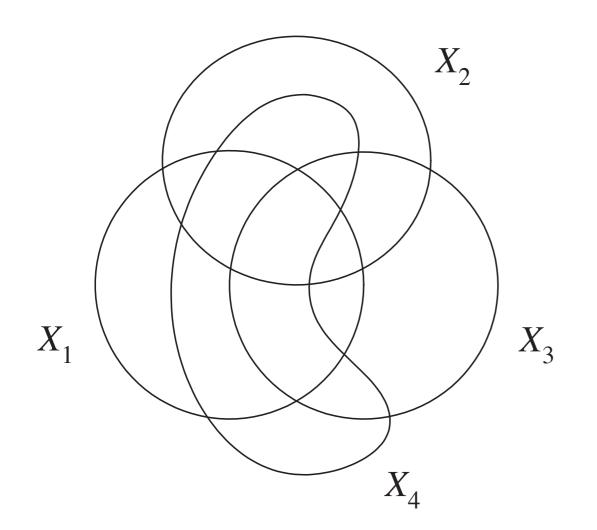




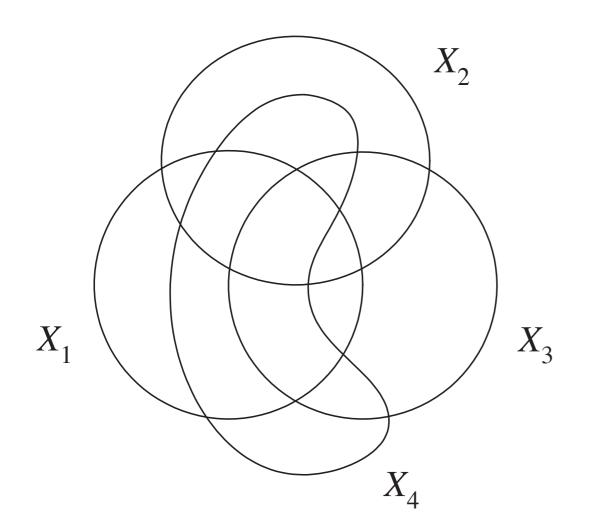




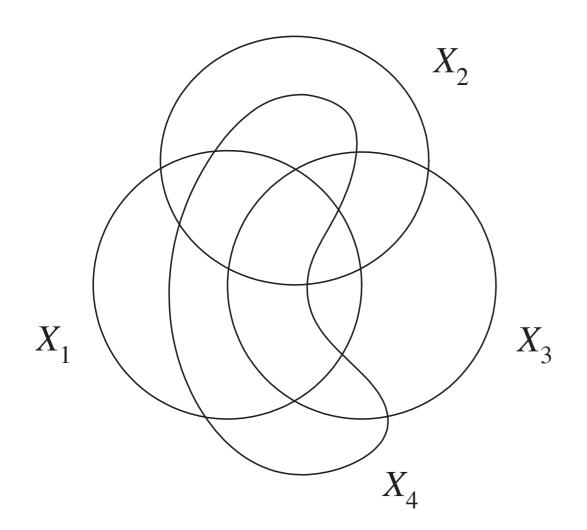
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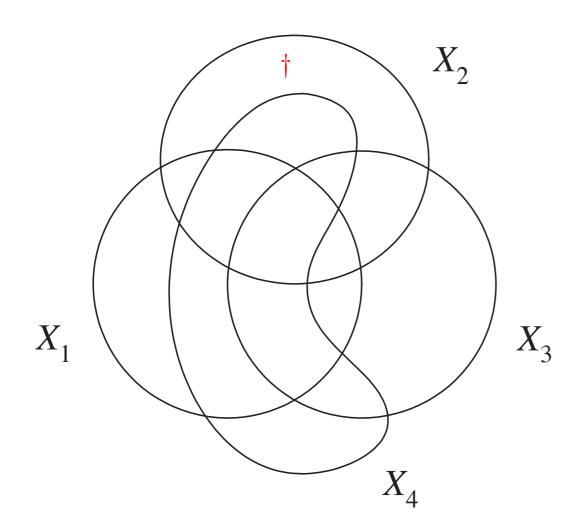


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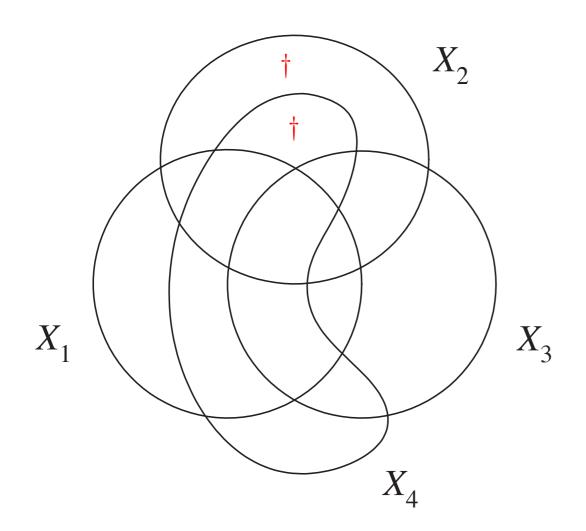
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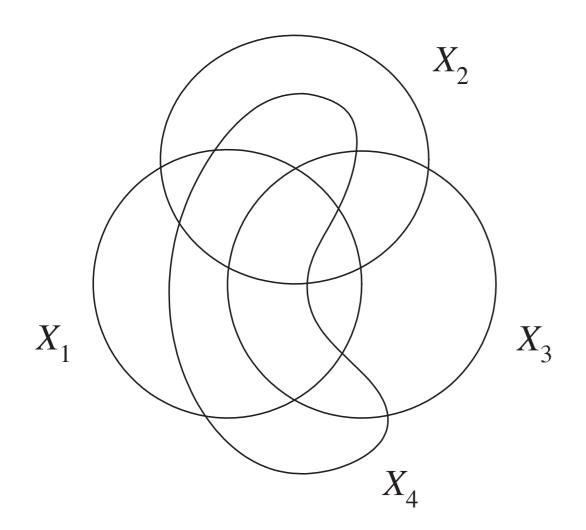
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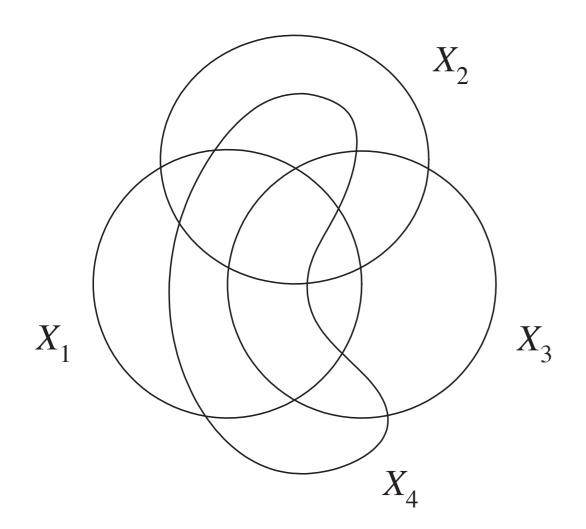
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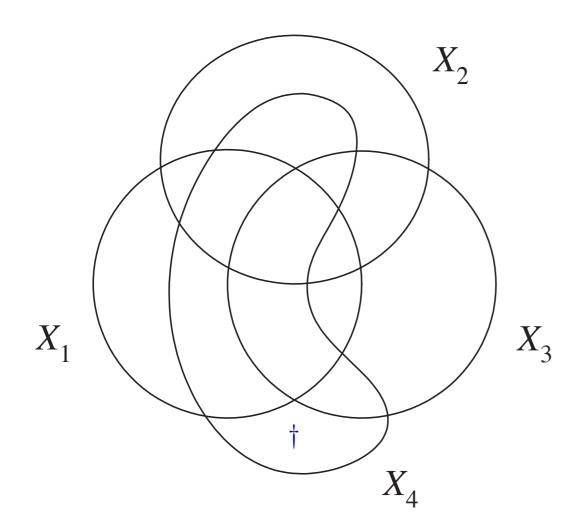
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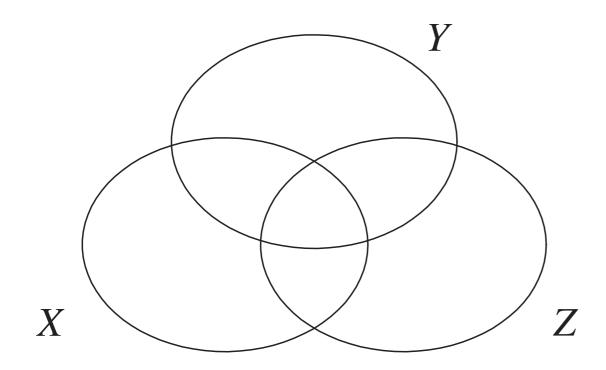
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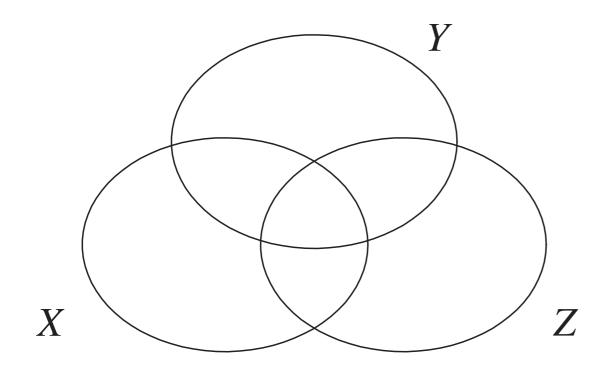
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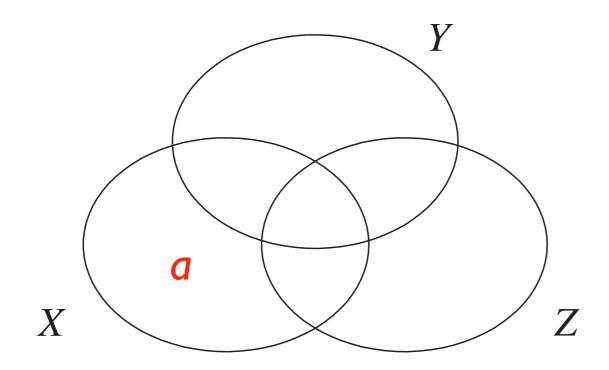
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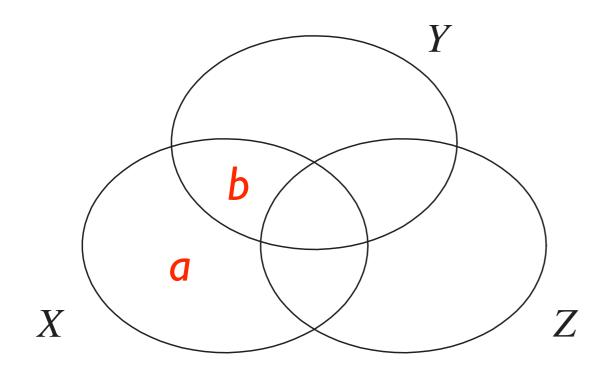
Idea



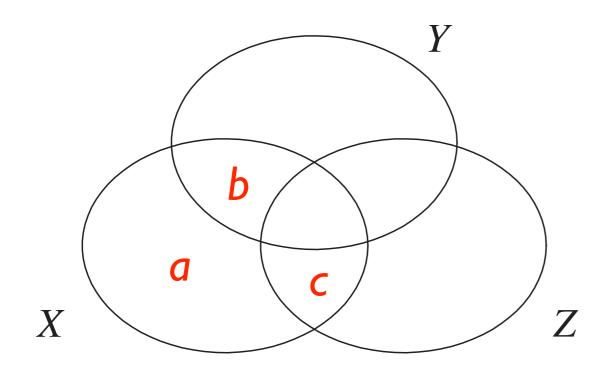
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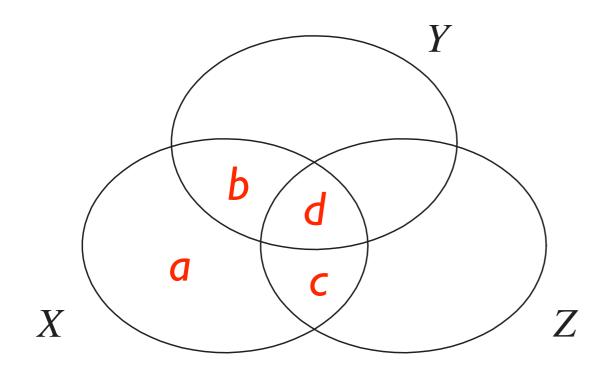
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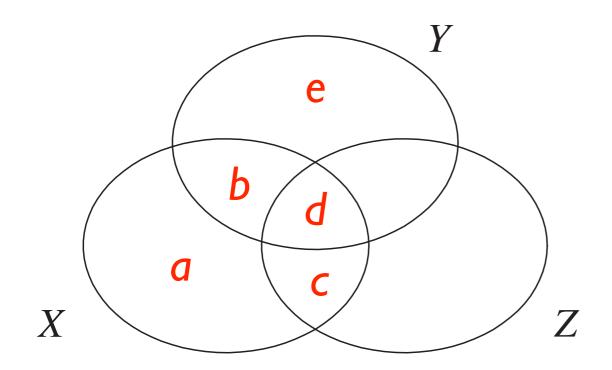
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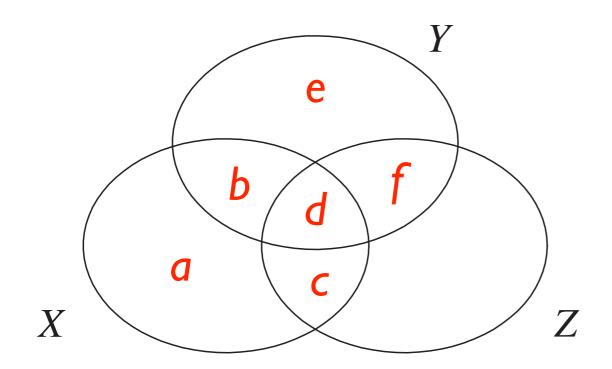
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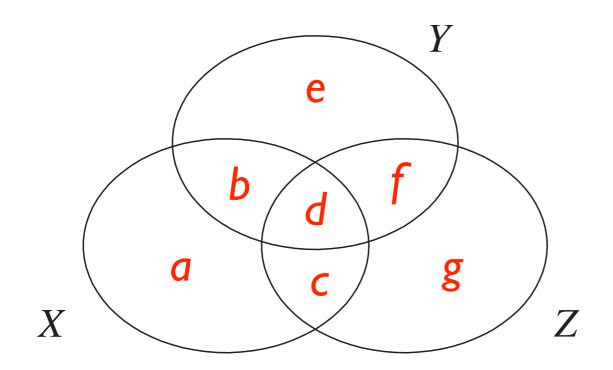
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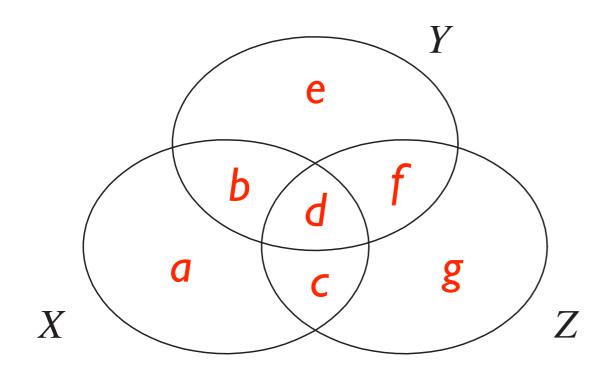


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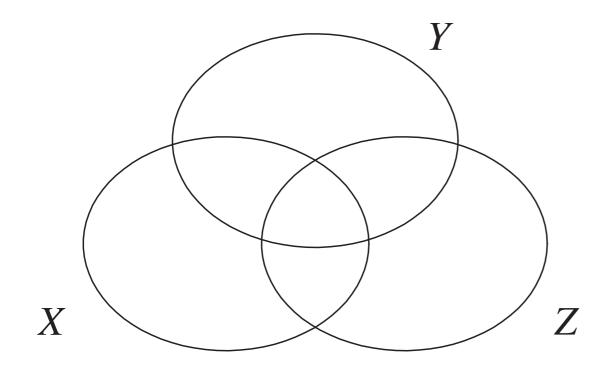


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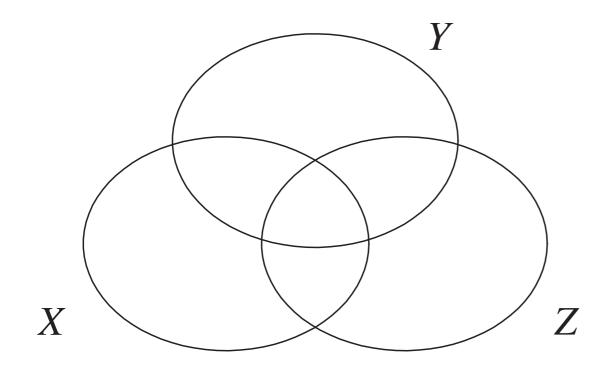




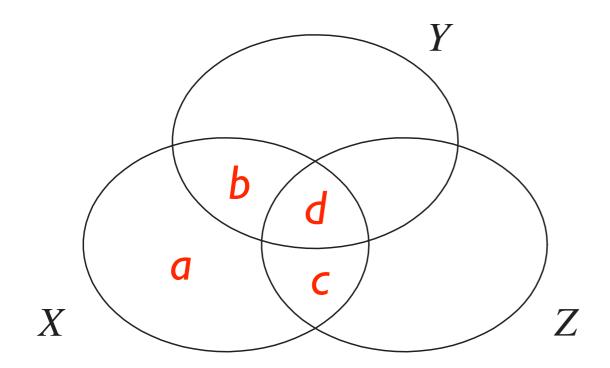
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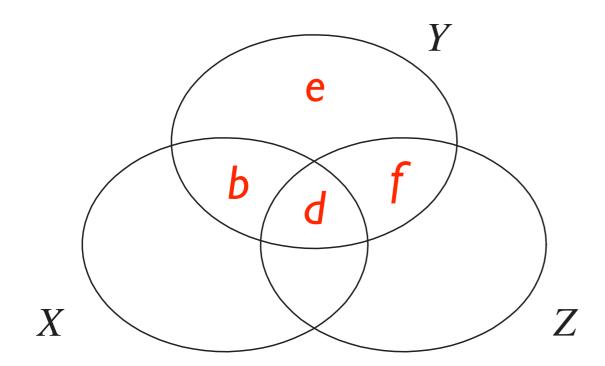


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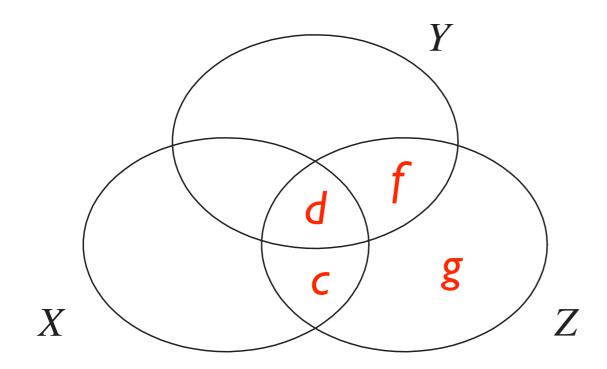
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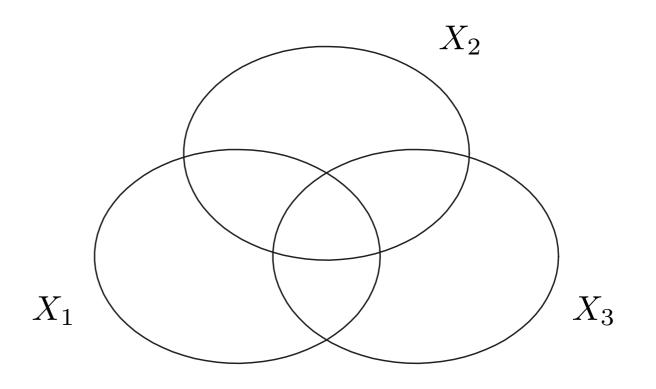
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**Corollary 2.44** The entropy of a random variable may take any nonnegative real value.

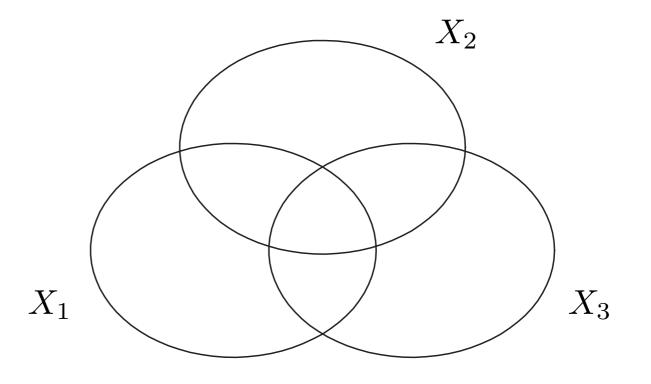
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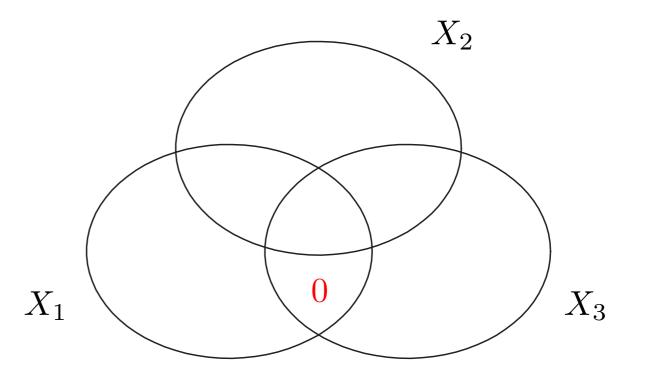
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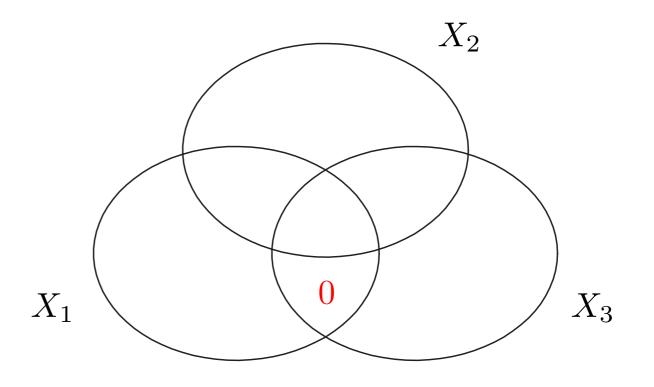
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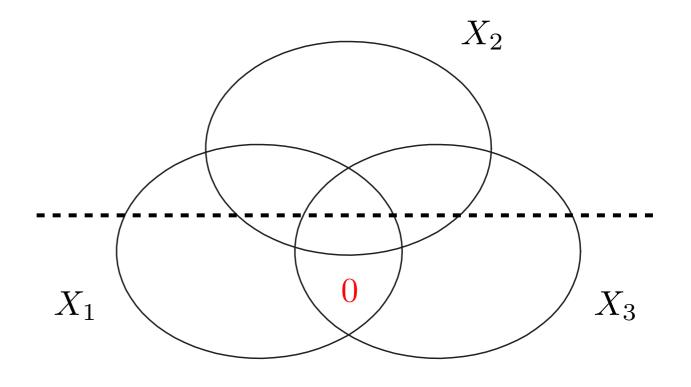
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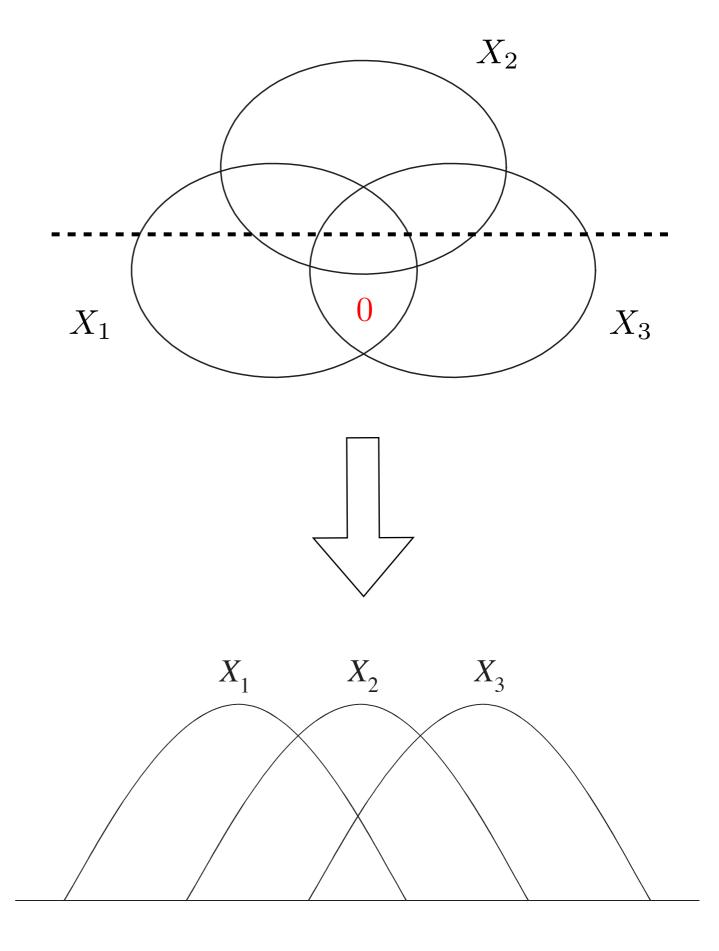
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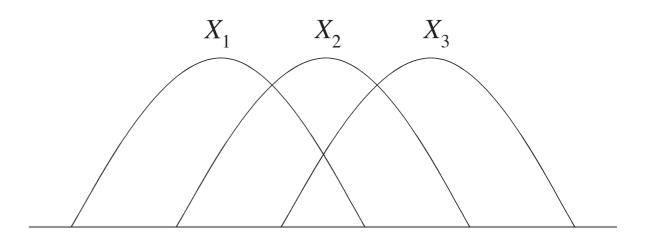
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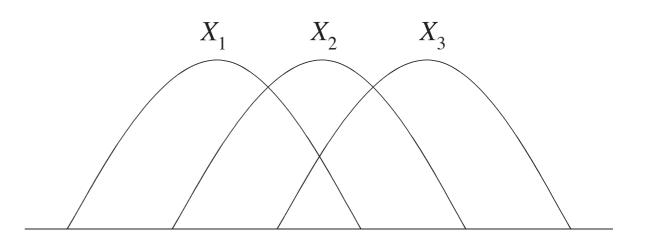


 $\mu^*$  for  $X_1 \to X_2 \to X_3$ 



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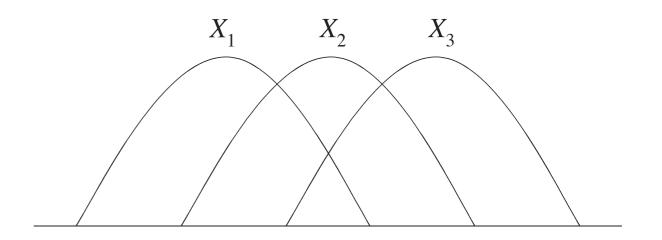
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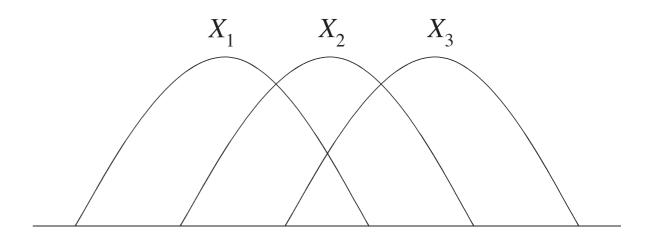
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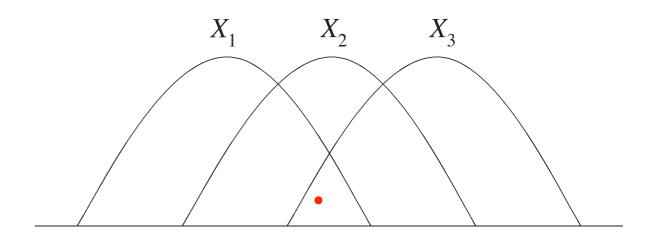
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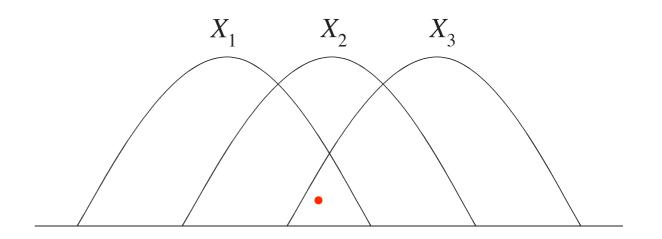
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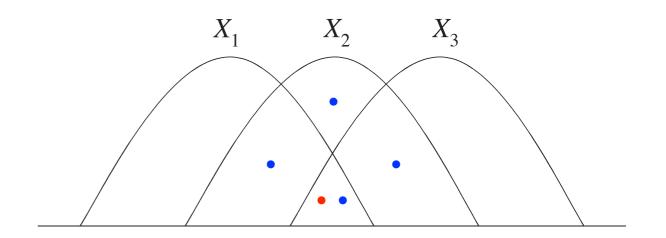
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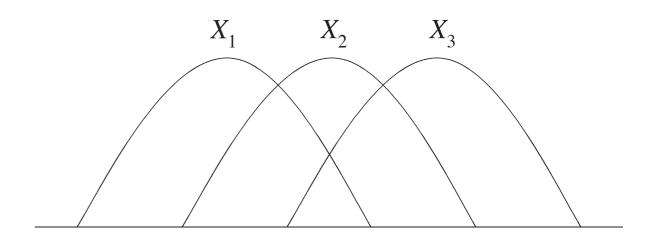
$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

$$I(X_1; X_3 | X_2) = \mu^* (\tilde{X}_1 \cap \tilde{X}_3 - \tilde{X}_2)$$



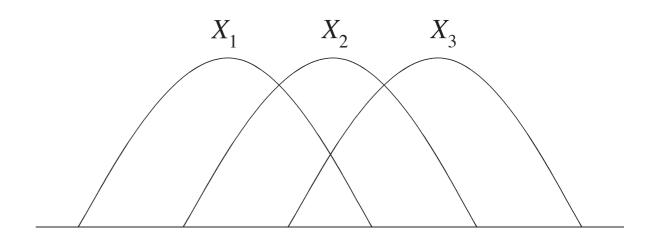
$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

$$I(X_1; X_3 | X_2) = \mu^* (\tilde{X}_1 \cap \tilde{X}_3 - \tilde{X}_2)$$
$$= \mu^* (\emptyset)$$



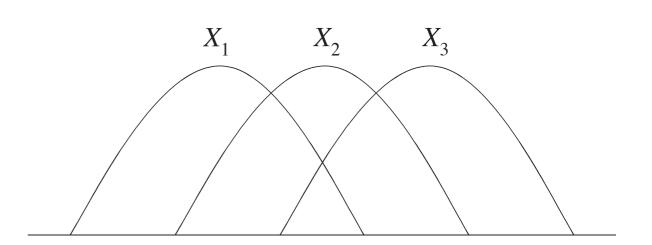
$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

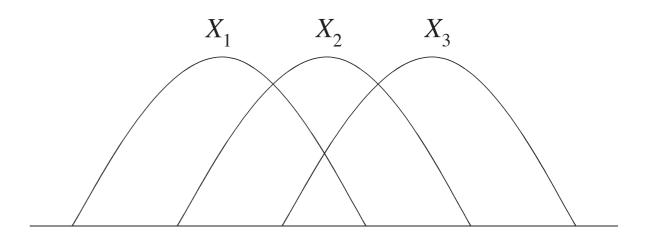
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.





$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

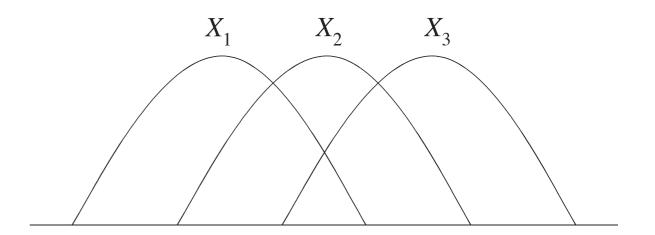
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3) = \mu^*(\tilde{X}_1 \cap \tilde{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

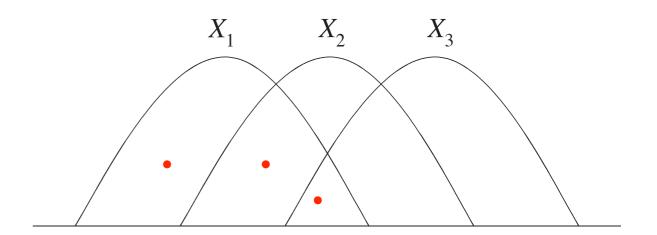
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \tilde{X}_2 \cap \tilde{X}_3) = \mu^*(\tilde{X}_1 \cap \tilde{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

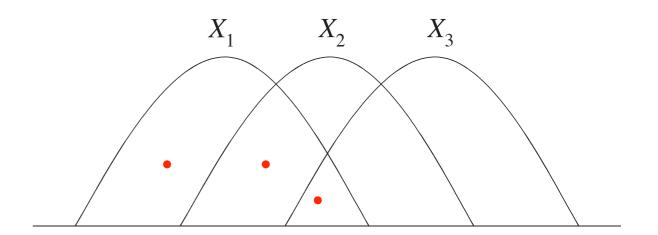
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_3})$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

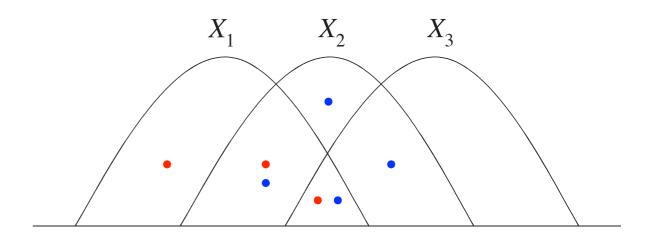
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3) = \mu^*(\tilde{X}_1 \cap \tilde{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

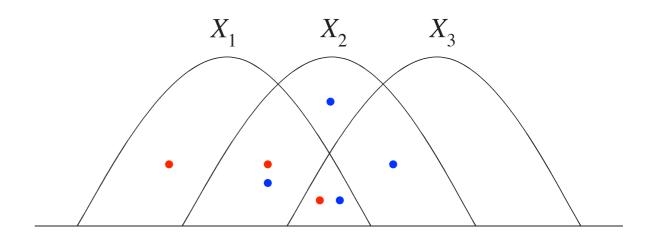
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3) = \mu^*(\tilde{X}_1 \cap \tilde{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

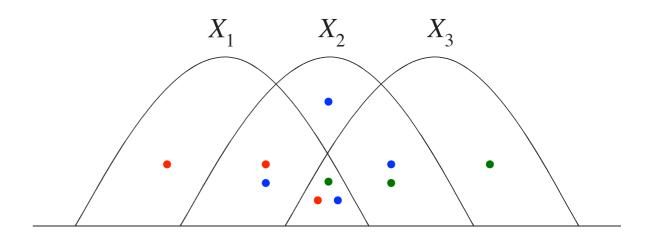
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\tilde{X}_1 \cap \bar{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

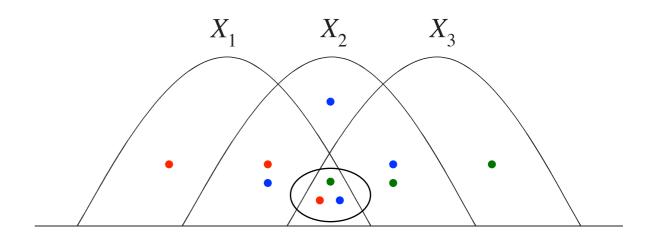
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\tilde{X}_1 \cap \bar{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

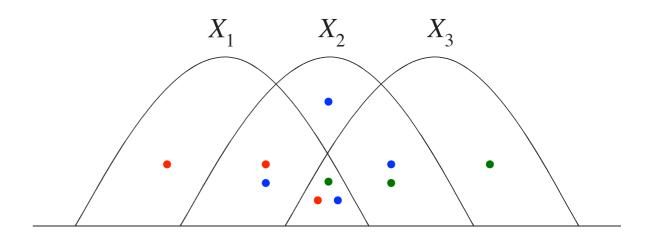
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
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$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\tilde{X}_1 \cap \overline{\tilde{X}_3})$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

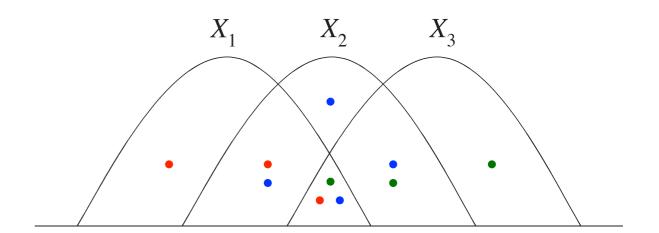
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\tilde{X}_1 \cap \bar{X}_3)$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

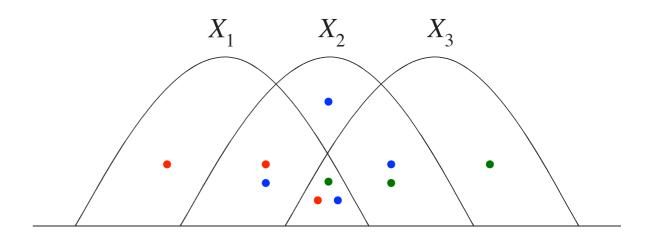
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
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= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\underline{\tilde{X}_1} \cap \overline{\tilde{X}_3})$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

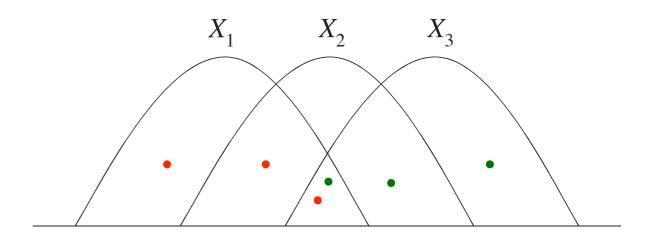
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_3})$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

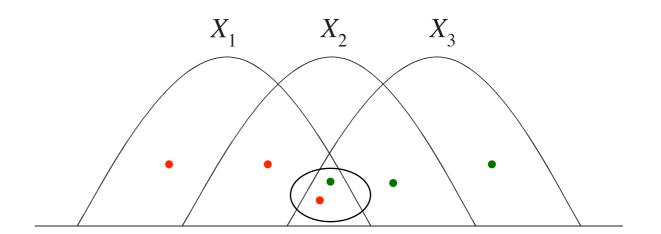
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_3})$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

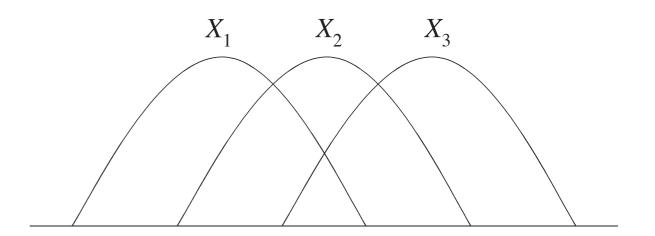
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_2} \cap \underline{\tilde{X}_3}) = \mu^*(\underline{\tilde{X}_1} \cap \underline{\tilde{X}_3})$$

$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

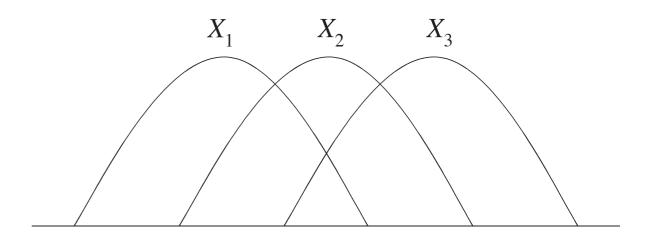
$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^* (\tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3) = \mu^* (\tilde{X}_1 \cap \tilde{X}_3) \\ = I(X_1; X_3)$$

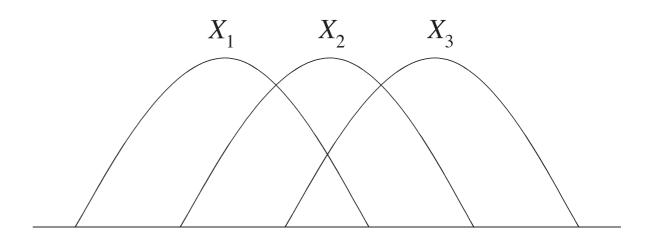
$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

$$I(X_{1}; X_{3} | X_{2}) = \mu^{*} (\tilde{X}_{1} \cap \tilde{X}_{3} - \tilde{X}_{2})$$
  
=  $\mu^{*} (\emptyset)$   
= 0.



$$\mu^*$$
 for  $X_1 \to X_2 \to X_3$ 

$$I(X_1; X_3 | X_2) = \mu^* (\tilde{X}_1 \cap \tilde{X}_3 - \tilde{X}_2)$$
$$= \mu^* (\emptyset)$$
$$= 0.$$



2. Also,

$$\mu^{*}(\tilde{X}_{1} \cap \tilde{X}_{2} \cap \tilde{X}_{3}) = \mu^{*}(\tilde{X}_{1} \cap \tilde{X}_{3})$$
$$= I(X_{1}; X_{3})$$
$$\geq 0.$$

3. Since the values of  $\mu^*$  on all the remaining atoms correspond to Shannon's information measures and hence are nonnegative, we conclude that  $\mu^*$  is a measure.

• For  $X_1 \to X_2 \to X_3 \to X_4$ ,  $\mu^*$  vanishes on the following 5 atoms:

$$\begin{split} \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3 \cap \tilde{X}_4^c \\ \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3 \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3^c \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3^c \cap \tilde{X}_4 \\ \tilde{X}_1^c \cap \tilde{X}_2 \cap \tilde{X}_3^c \cap \tilde{X}_4 \end{split}$$

• For  $X_1 \to X_2 \to X_3 \to X_4$ ,  $\mu^*$  vanishes on the following 5 atoms:

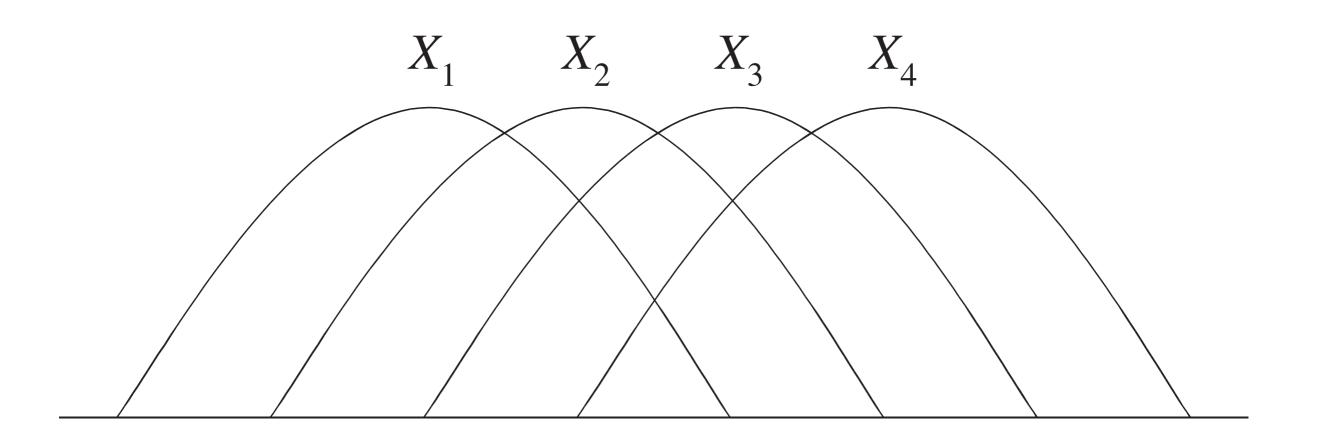
$$\begin{split} \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3 \cap \tilde{X}_4^c \\ \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3 \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3^c \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3^c \cap \tilde{X}_4 \\ \tilde{X}_1^c \cap \tilde{X}_2 \cap \tilde{X}_3^c \cap \tilde{X}_4 \end{split}$$

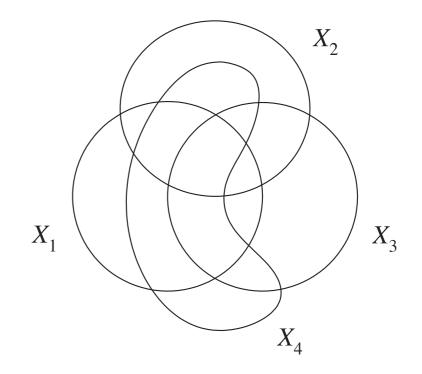
• The information diagram can be displayed in two dimensions.

• For  $X_1 \to X_2 \to X_3 \to X_4$ ,  $\mu^*$  vanishes on the following 5 atoms:

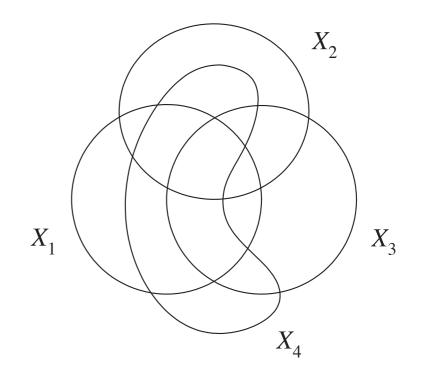
$$\begin{split} \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3 \cap \tilde{X}_4^c \\ \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3 \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2^c \cap \tilde{X}_3^c \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3^c \cap \tilde{X}_4 \\ \tilde{X}_1 \cap \tilde{X}_2 \cap \tilde{X}_3^c \cap \tilde{X}_4 \end{split}$$

- The information diagram can be displayed in two dimensions.
- The values of  $\mu^*$  on the remaining atoms correspond to Shannon's information measures and hence are nonnegative. Thus,  $\mu^*$  is a measure.



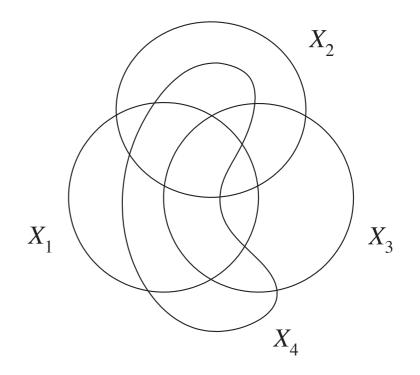


1. The Markov subchain  $X_1 \to X_2 \to X_3$  implies



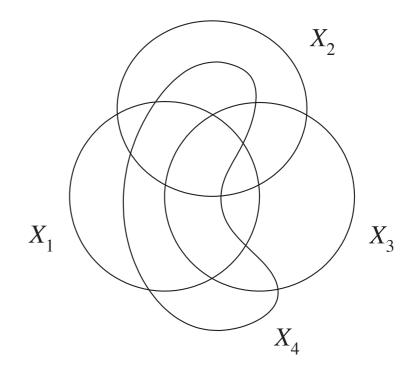
1. The Markov subchain  $X_1 \to X_2 \to X_3$  implies

 $0 = I(X_1; X_3 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_3 | X_2, X_4).$ 



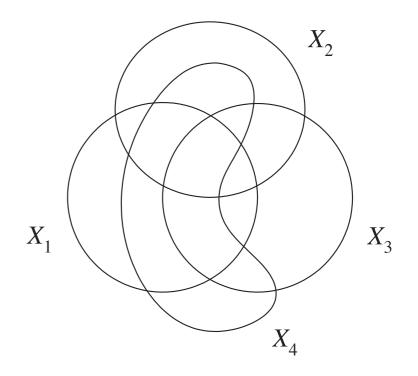
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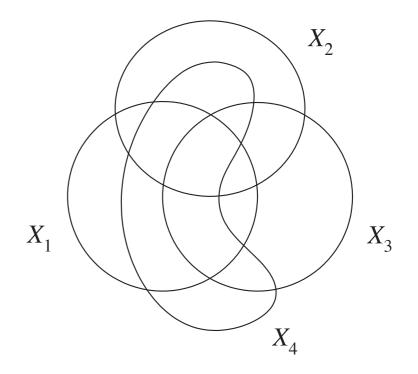
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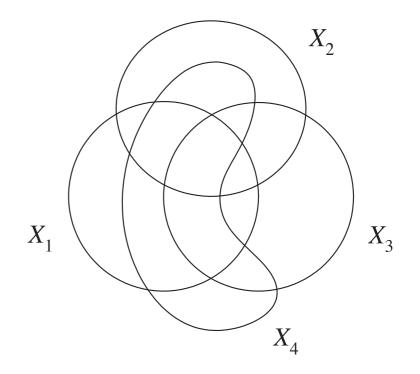
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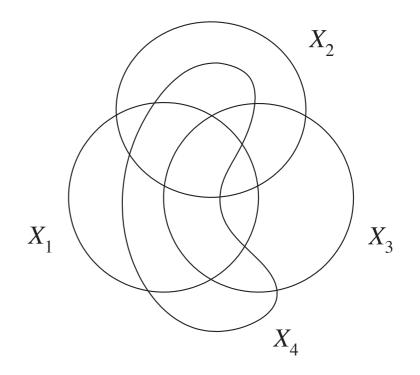
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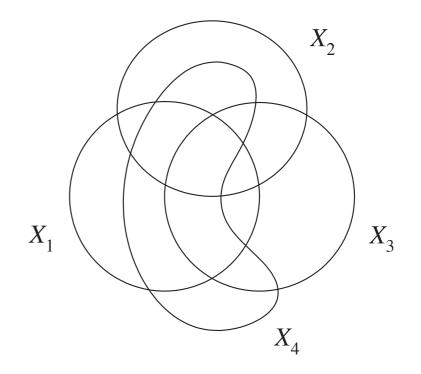
 $0 = I(X_1; X_3 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_3 | X_2, X_4).$ 



1. The Markov subchain  $X_1 \to X_2 \to X_3$  implies

 $0 = I(X_1; X_3 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_3 | X_2, X_4).$ 

Let  $I(X_1; X_3 | X_2, X_4) = a \ge 0$ . Then

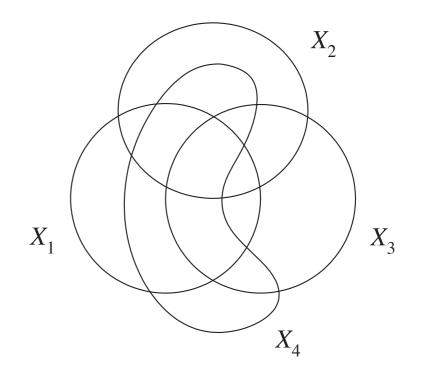


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Let  $I(X_1; X_3 | X_2, X_4) = a \ge 0$ . Then

 $\underline{I(X_1;X_3;X_4|X_2) = -a}.$ 

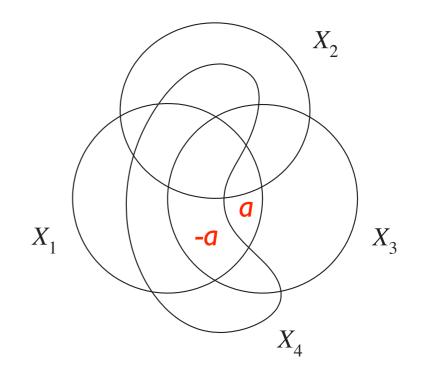


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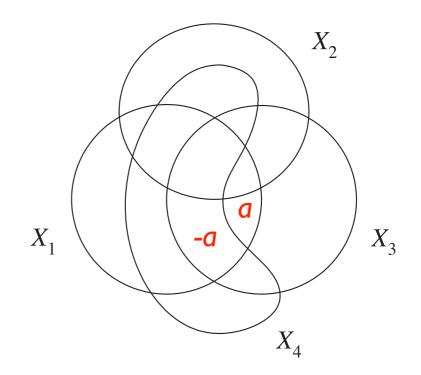


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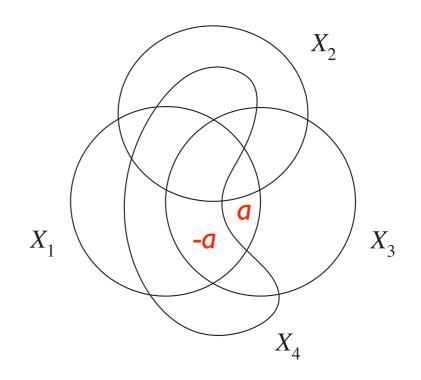


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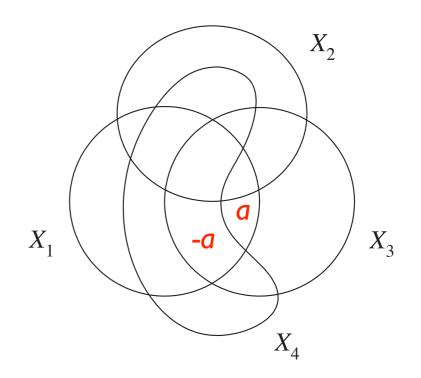
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Let  $I(X_1; X_3 | X_2, X_4) = a \ge 0$ . Then

$$I(X_1; X_3; X_4 | X_2) = -a.$$

- 2. The Markov subchain  $X_1 \to X_2 \to X_4$  implies
- $0 = I(X_1; X_4 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_4 | X_2, X_3).$



1. The Markov subchain  $X_1 \to X_2 \to X_3$  implies

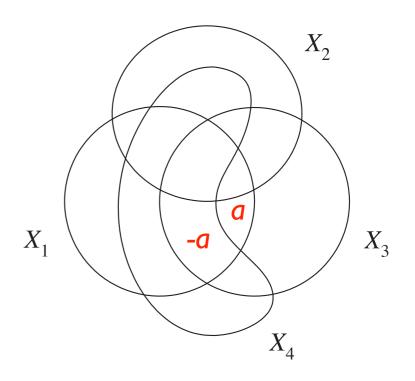
$$0 = I(X_1; X_3 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_3 | X_2, X_4).$$

Let  $I(X_1; X_3 | X_2, X_4) = a \ge 0$ . Then

 $I(X_1; X_3; X_4 | X_2) = -a.$ 

2. The Markov subchain  $X_1 \to X_2 \to X_4$  implies

 $0 = I(X_1; X_4 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_4 | X_2, X_3).$ 



1. The Markov subchain  $X_1 \to X_2 \to X_3$  implies

$$0 = I(X_1; X_3 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_3 | X_2, X_4).$$

Let  $I(X_1; X_3 | X_2, X_4) = a \ge 0$ . Then

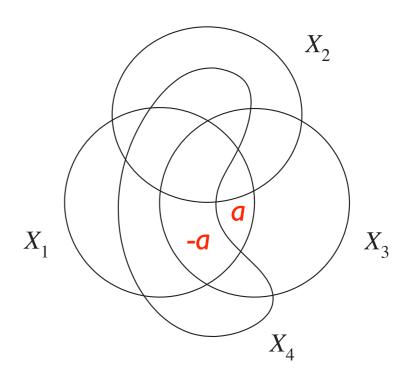
 $I(X_1; X_3; X_4 | X_2) = -a.$ 

2. The Markov subchain  $X_1 \to X_2 \to X_4$  implies

 $0 = I(X_1; X_4 | X_2) = I(X_1; X_3; X_4 | X_2) + I(X_1; X_4 | X_2, X_3).$ 

Since  $I(X_1; X_3; X_4 | X_2) = -a$ ,

 $I(X_1; X_4 | X_2, X_3) = a.$ 



1. The Markov subchain  $X_1 \to X_2 \to X_3$  implies

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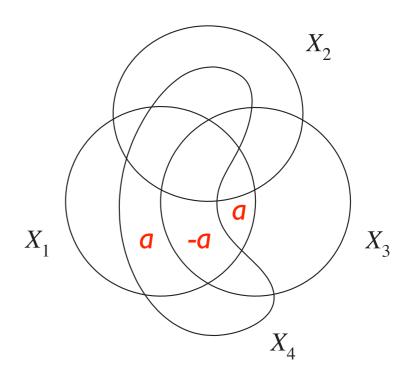
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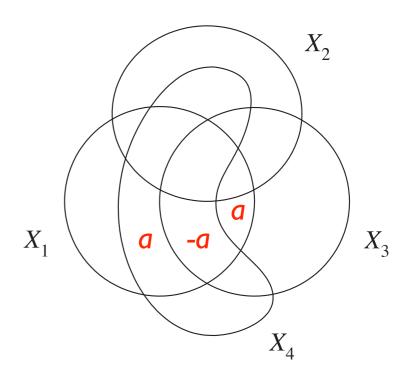
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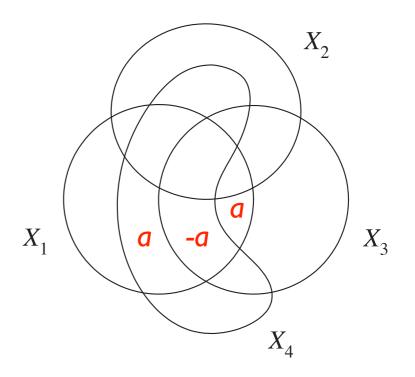
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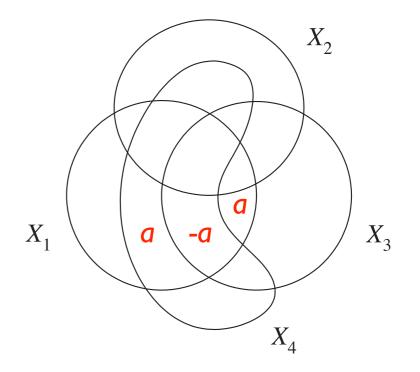
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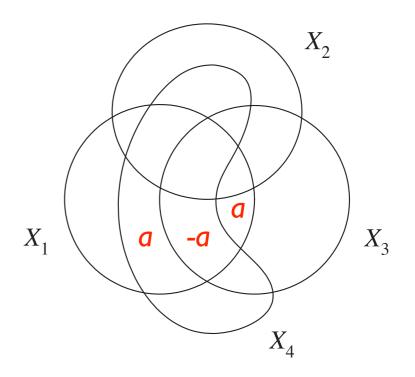
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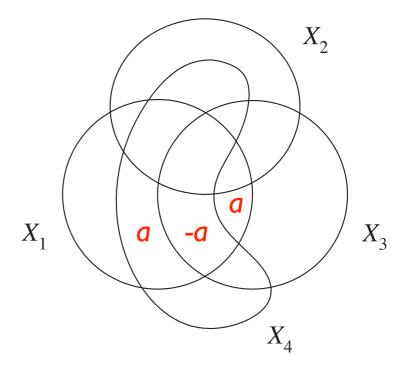
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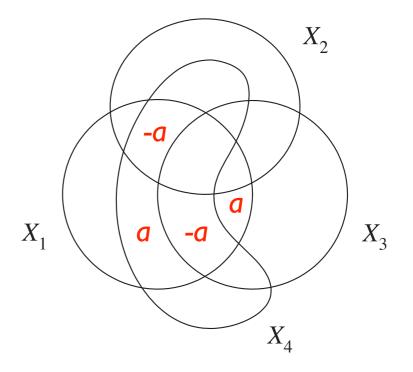
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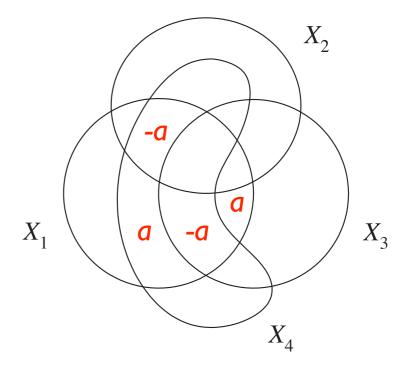
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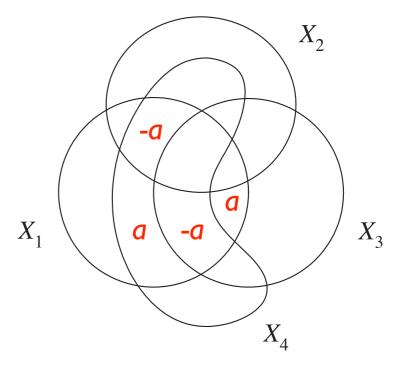
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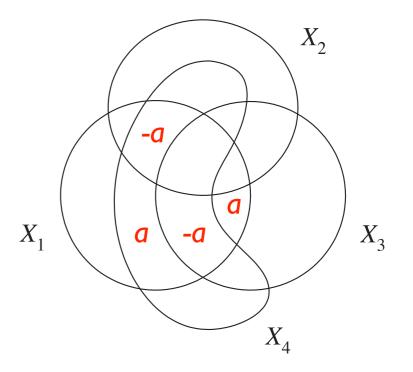
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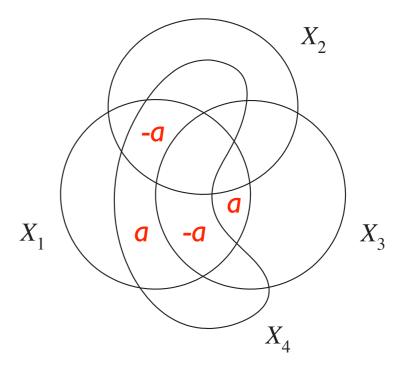
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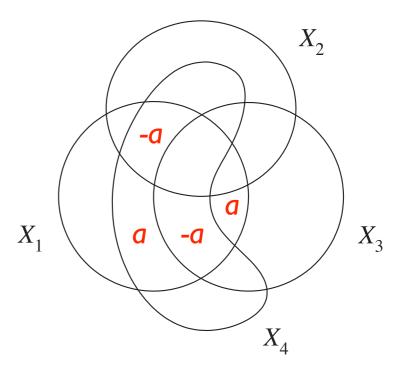
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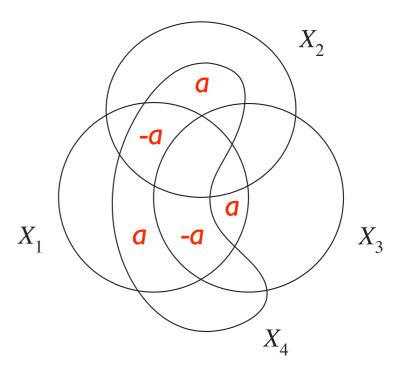
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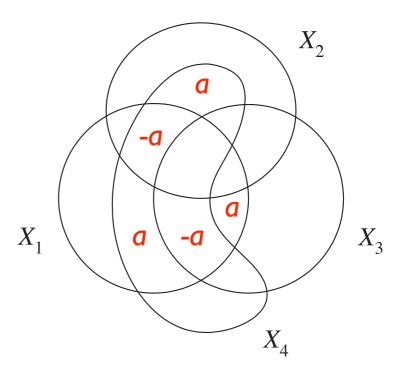
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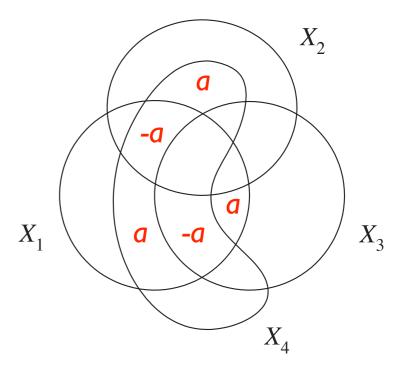
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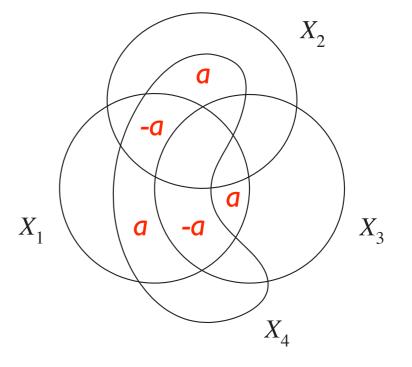
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Since  $I(X_1; X_3; X_4 | X_2) = -a$ ,

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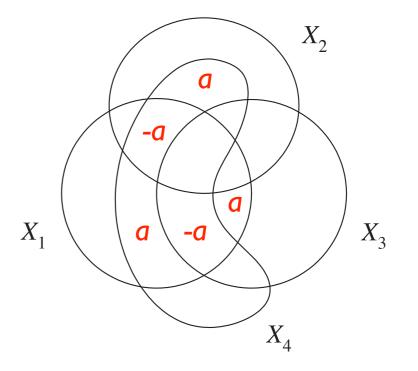
- $I(X_1; X_2; X_4 | X_3) = -a.$
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$$\begin{array}{rcl} 0 &=& I(X_1,X_2;X_4|X_3) &= \\ & I(X_1;X_4|X_2,X_3) + I(X_1;X_2;X_4|X_3) + I(X_2;X_4|X_1,X_3) \end{array}$$



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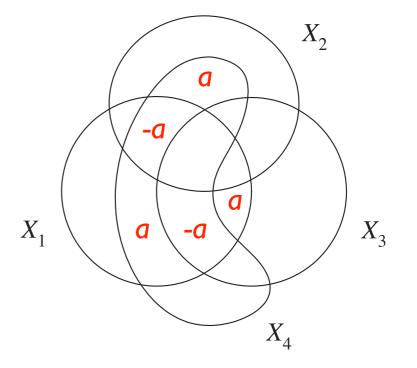
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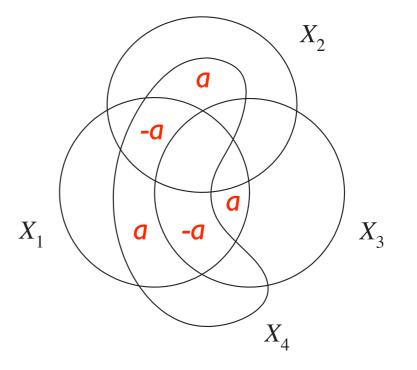
$$0 = I(X_2; X_4 | X_3) = I(X_1; X_2; X_4 | X_3) + I(X_2; X_4 | X_1, X_3)$$

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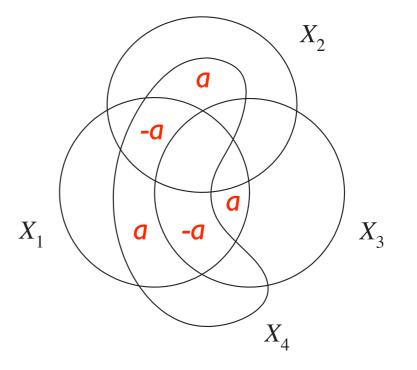
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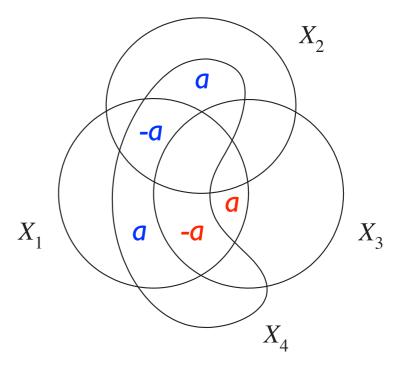
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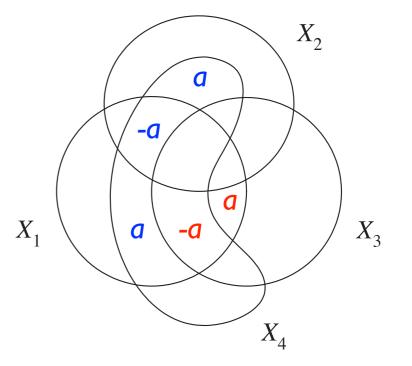
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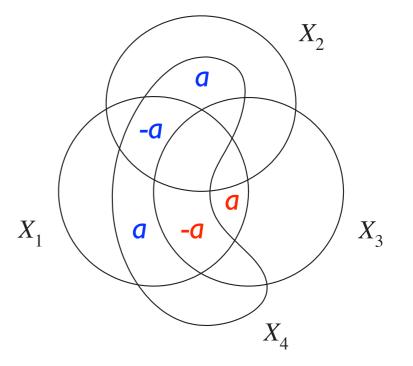
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Therefore a = 0, and so  $\mu^*$  vanishes on the corresponding 5 atoms as shown in the information diagram.



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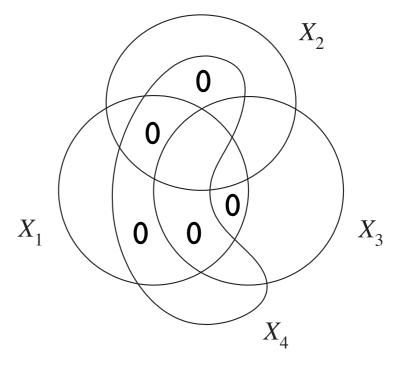
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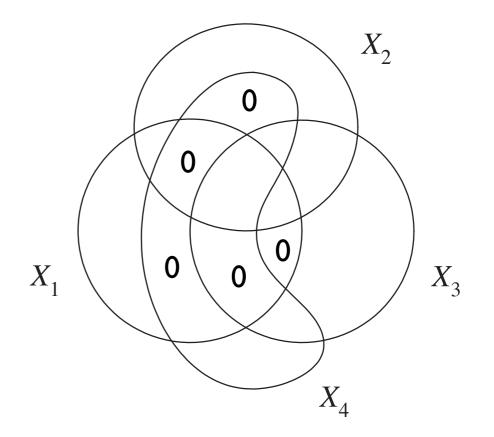
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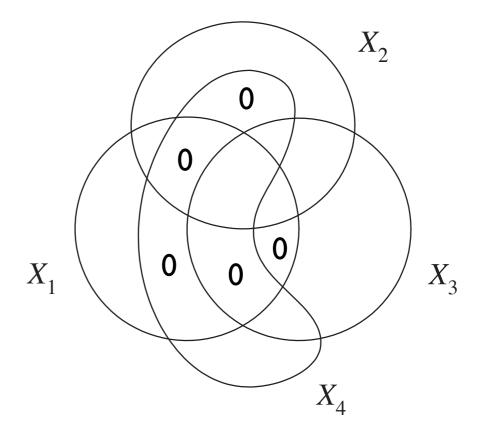
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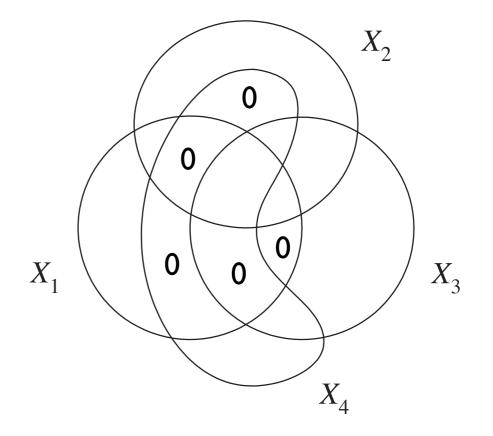


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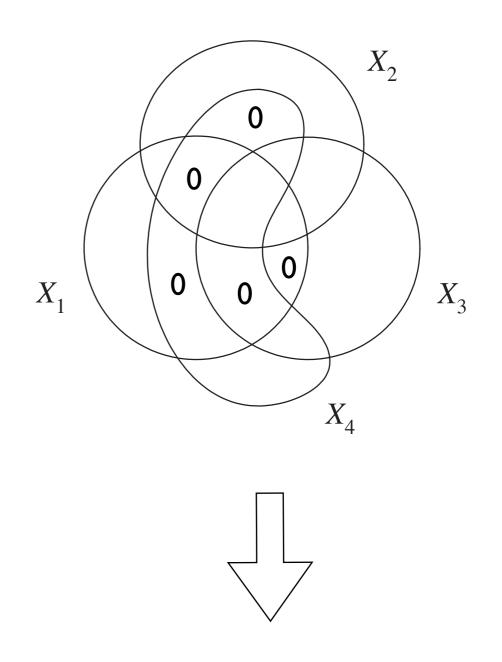
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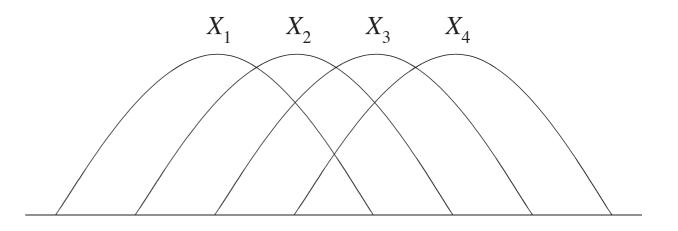
2. Suppress these atoms by setting them to  $\emptyset$  to obtain the information diagram below.



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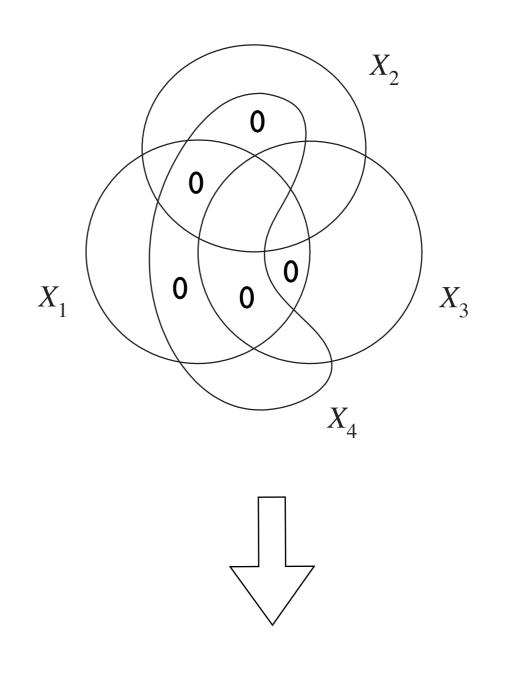


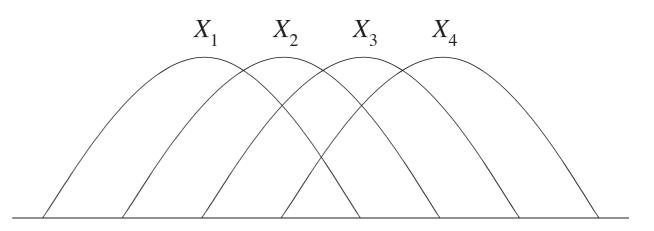


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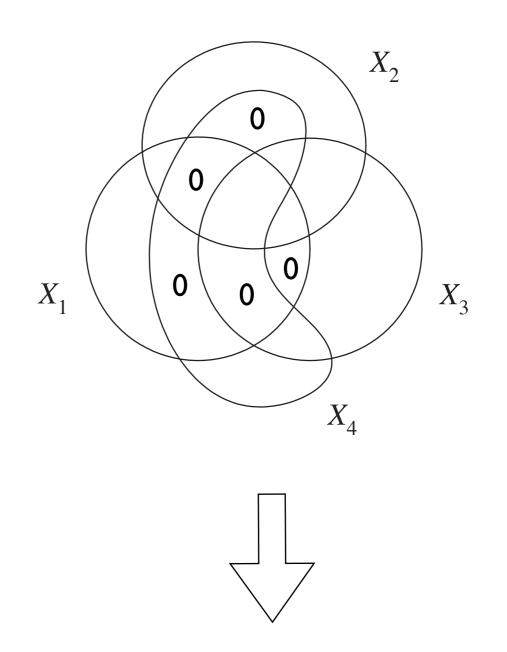
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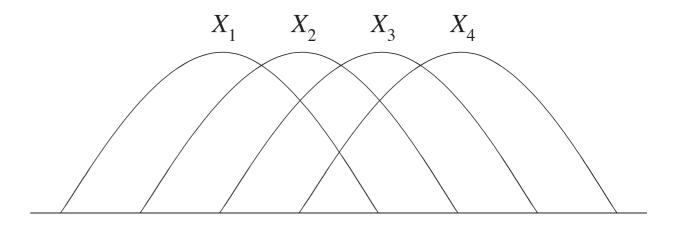
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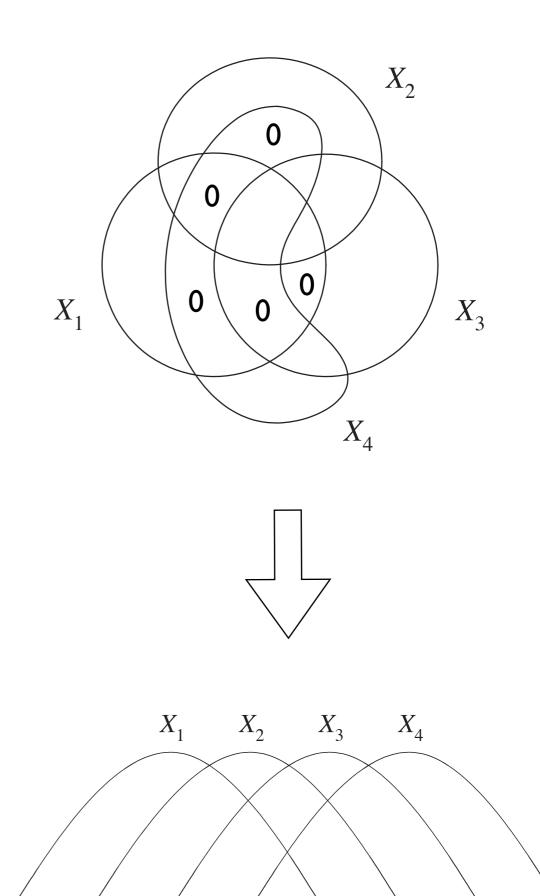
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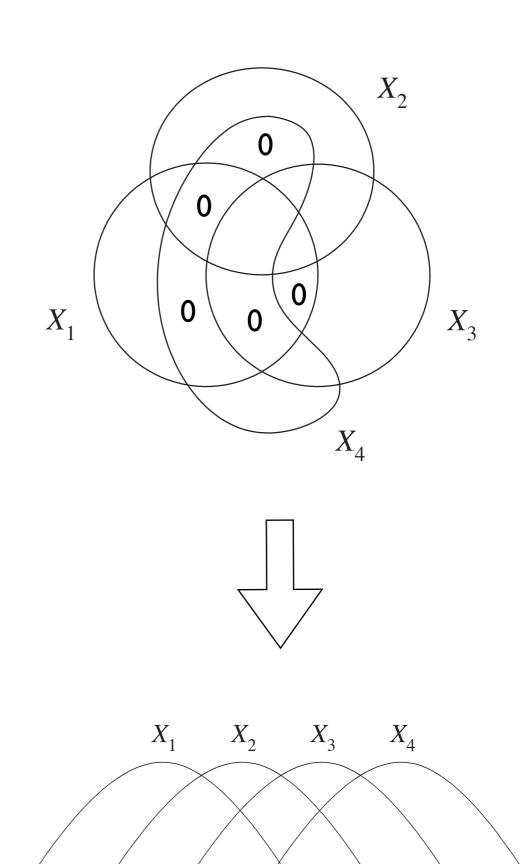
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3. There are all Shannon's information measures which are always nonnegative. Therefore,  $\mu^*$  is a measure.



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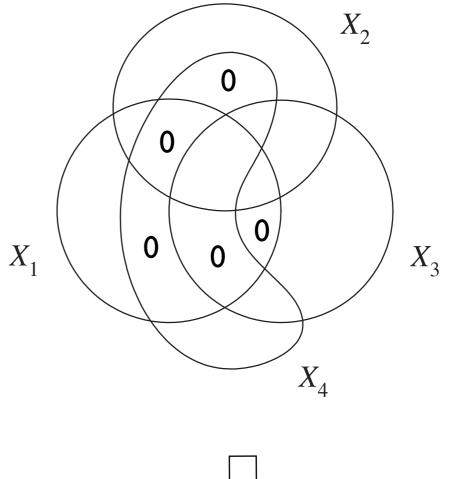
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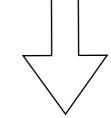
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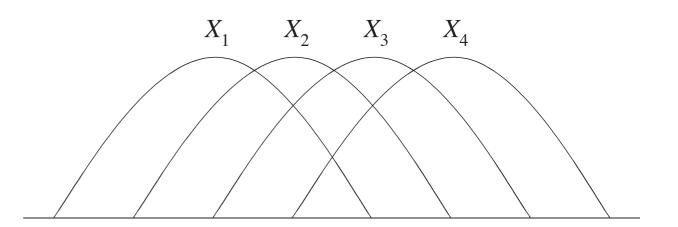
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Exercise: Identify these 10 atoms in the information diagram at the bottom.

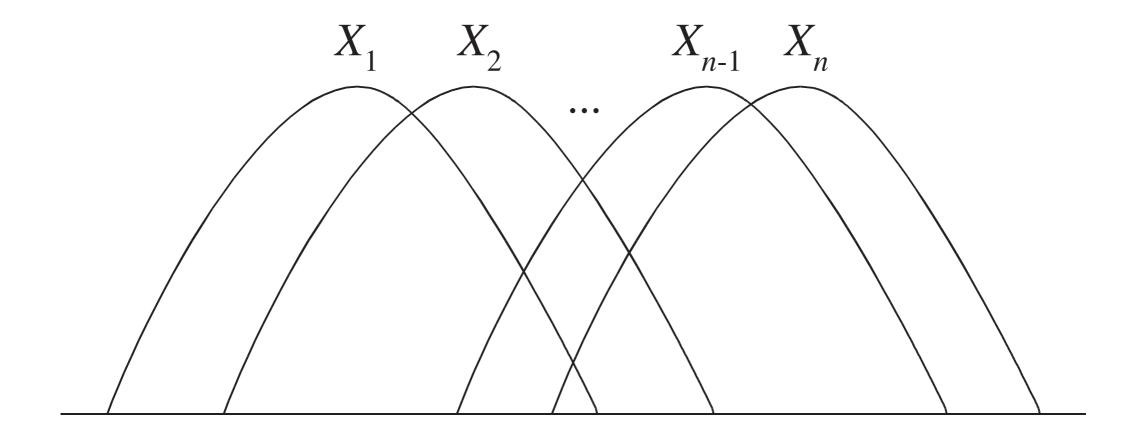




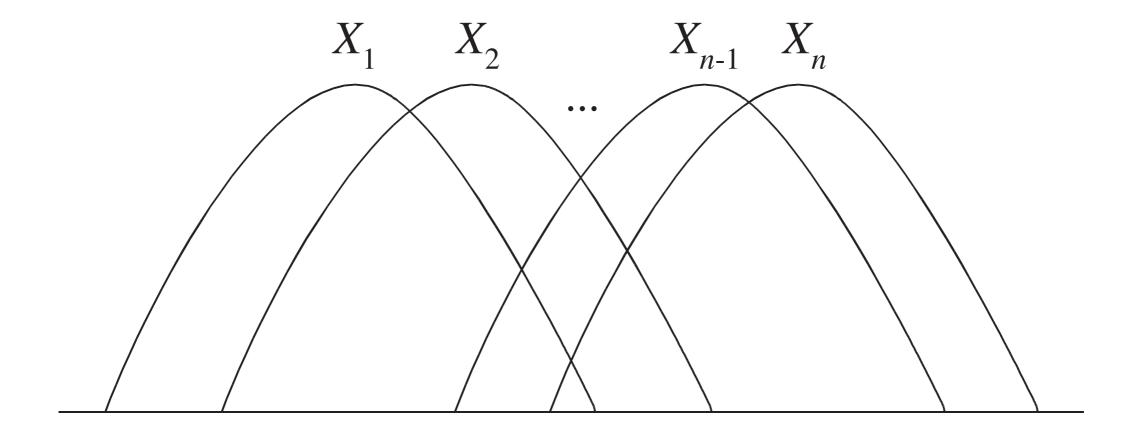


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- See Ch. 12 for a detailed discussion in the context of Markov random field.

