

CMPSC 165B: Homework 4

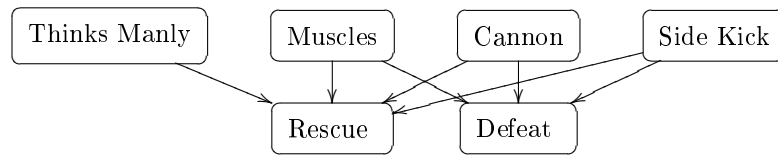
Due on June 3, 2011

James Schaffer: Monday 5:00p

Kyle Ibrahim

Contents

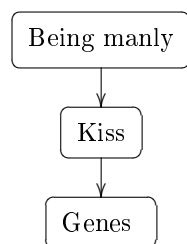
Problem 1 (Bayesian Belief Networks)	2
Problem 2 (Conditional Independence)	3
Problem 3 (Learning Theory)	3
Problem 4 (Regression)	4

Problem 1 (Bayesian Belief Networks)

Arc values:

P(Muscles)		P(Cannon)		P(Side Kick)		P(Thinks Manly)	
wimpy	.33	big	.33	comic relief	.5	T	.5
average	.33	gigantic	.33	loyal martyr	.5	F	.5
ripped	.34	obscene	.34				

Note, whether or not Sally thinks you're manly is independent of being manly, thus, we don't need more than the above arcs for the above network.



Arc values:

P(Being Manly)		Kiss	Being Manly	P(Kiss)	Genes	Kiss	P(Genes)
T	.5	T	T	.5	T	T	.5
F	.5	T	F	.5	T	F	.5
		F	T	.5	F	T	.5
		F	F	.5	F	F	.5

1. As these are independent: $P(\text{BeingManly}=T) = .5$
2. Again, these are independent: $P(\text{BeingManly}=T) = .5$
3. $.5 * .5P(\text{Being Manly}=T) + .5 * .5P(\text{Being Manly}=F) = .25 + .25 = .5$
4. $.5 * .5P(\text{Kiss}=T) + .5 * .5P(\text{Kiss}=F) = .25 + .25 = .5$
5. $.5 * .5P(\text{Kiss}=T) + .5 * .5P(\text{Kiss}=F) = .25 + .25 = .5$, as this is independent of rescuing Sally.

Problem 2 (Conditional Independence)

1. Yes, as they have no common ancestors and Z is not blocking
2. No, as W is a direct decedent of X and Z is not blocking
3. No, as Y remains a common ancestor
4. No, as Y remains a common ancestor
5. Yes given nothing else as they have no common ancestors
6. Yes given nothing else as there is not path from X to Z

Problem 3 (Learning Theory)

Assuming “better” infers more accurate classification in the average case, then no: a naive Bayesian classifier will, in general, perform as well as other classifiers over truly independent attributes.

Problem 4 (Regression)

1. It is a non-linear regression problem because w , which is the unknown to be solved, cannot be expressed as a linear function of \mathbf{x} .
2. We can use the log likelihood to find \hat{w} :

$$\begin{aligned}
 \ln \mathcal{L}(w|x_i) &= \ln \left(\prod_{i=1}^n f(x_i|w) \right) \\
 &= \sum_{i=1}^n \ln \left(\frac{1}{\sqrt{2\pi}} \exp \left(-\frac{(y_i - e^{wx_i})^2}{2} \right) \right) \\
 &= \sum_{i=1}^n \ln \left(\frac{1}{\sqrt{2\pi}} \right) - \frac{(y_i - e^{wx_i})^2}{2}, \quad \hat{\ell} = \frac{\ln \mathcal{L}}{n}
 \end{aligned}$$

$$\begin{aligned}
 \frac{\partial \ln \mathcal{L}}{\partial w} &= 0 \Rightarrow \\
 \frac{\sum_{i=1}^n \ln \left(\frac{1}{\sqrt{2\pi}} \right) - \frac{(y_i - e^{wx_i})^2}{2}}{\partial w} &= 0 \Rightarrow \\
 \frac{n \ln \left(\frac{1}{\sqrt{2\pi}} \right) - \sum_{i=1}^n \frac{y_i^2}{2} - y_i e^{wx_i} + \frac{(e^{wx_i})^2}{2}}{\partial w} &= 0 \Rightarrow \\
 -\sum_{i=1}^n -x_i y_i e^{wx_i} + x_i e^{2wx_i} &= 0 \Rightarrow \\
 \sum_{i=1}^n x_i e^{2wx_i} &= \sum_{i=1}^n x_i y_i e^{wx_i}
 \end{aligned}$$

Thus, equation ii should be used.