**T.61.5140 Machine Learning: Advanced Probablistic Methods** Hollmén, Raiko (Spring 2008) Problem session, 25th of January, 2008 http://www.cis.hut.fi/Opinnot/T-61.5140/

1. Consider a bent coin and how to estimate the probability of tails  $\mu$ . The random variable  $X \in \{0,1\}$  (heads=0, tails=1) is distributed accoring to the Bernoulli distribution with the parameter  $\mu$  (see page 685 in Bishop, 2006).

(a) Derive a maximum likelihood estimator for  $\mu$  and estimate  $\hat{\mu}$  for the data set from the lecture (7 heads and 5 tails out of 12 tosses).

(b) Using a fair coin, what is the probability that out of 12 tosses, strictly more than 10 are heads (see Binomial distribution, page 686).

2. Compute the probability  $P(C \mid X)$  of using each coin in the guessing game from the lecture (see Bayes' theorem, p. 15). There are two bent coins ( $C \in \{c_1, c_2\}$ ) with different properties and the player guesses which coin was used after learning whether the toss was head or tails. The properties of the coins are:  $P(X = t \mid C = c_1) = \theta_1$  and  $P(X = t \mid C = c_2) = \theta_2$ . The used coin is chosen randomly by  $P(C = c_1) = \pi_1$  and  $P(C = c_2) = \pi_2$  with  $\pi_1 + \pi_2 = 1$ .

3. The Naïve Bayes model has a class label *C* and observations  $X_1, X_2, ..., X_6$  such that  $P(X_1, X_2, X_3, X_4, X_5, X_6, C) = P(C)P(X_1|C)P(X_2|C)...P(X_6|C)$ . (a) Simplify  $P(X_1 | C, X_2)$ 

(b) Solve the classification problem:  $P(C \mid X_1, X_2, ..., X_6)$ 

4. Draw a graphical representation of the models in problems 1, 2, and 3 where nodes represent random variables and arrows represent direct dependencies (see Bayesian Networks, page 360).