**T.61.5140 Machine Learning: Advanced Probablistic Methods** Hollmén, Raiko (Spring 2008) Problem session, 4th and 11th of April, 2008 http://www.cis.hut.fi/Opinnot/T-61.5140/

These are programming excercises that are discussed on 4th of April and demonstrated on 11th of April. The Matlab files are on the course webpage.

Let us use a notation  $N(x \mid \mu, \sigma^2)$  to denote a Gaussian distribution of x with mean  $\mu$  and variance  $\sigma^2$ .

1. Study the unknown distribution  $p_u(x)$  by drawing samples from it. You can get values of  $p_u(x)$  by calling the Matlab function unknown\_ p.m with the argument x which may be a scalar or a vector. Use rejection sampling (page 528) with a proposal distribution q(x) = N(0,2) (Gaussian with mean 0 and variance 2) and scaling k=1.5 (You get values of kq(x) as 1.5\*gaussian(x,0,2)). (a) Plot a figure that shows the accepted samples with a black dot ('.k') and rejected samples with a cyan dot ('.c'). Use both 100 and 1000 samples. (b) Estimate the expected values E(x) and E(tanh(x))over distribution  $p_u(x)$  by using the accepted samples.

2. Study the same distribution  $p_u(x)$  by using importance sampling (page 532). (a) Plot the importance weights  $p_u(x)/q(x)$ . (b) Estimate E(x) and  $E(\tanh(x))$  over distribution  $p_u(x)$  by weighting all samples. (c) Was it more accurate than rejection sampling?

3. Let us study a model with two variables:  $x_1 \sim N(x_1 \mid 0, 1.0)$  and  $x_2 \sim N(x_2 \mid x_1, 0.1)$  where the two Gaussians are independent. (a) Solve  $p(x_1 \mid x_2)$ . (b) Initialize  $x_1^{(0)} = x_2^{(0)} = 0$  and plot 50 samples from the model by Gibbs sampling (page 542) with a line connecting consequtive samples.

4. Let us study a model with three variables,  $x_1$ ,  $x_2$ , and  $x_3$ . The model is such that  $x_1$  and  $x_2$  are drawn from the same distribution  $p_u(x)$  as before, and  $x_3 \sim N(x_3 \mid x_1 + x_2, 1.0)$ . Draw samples from  $p(x_1, x_2 \mid x_3 = 1.0)$  by using Metropolis algorithm (page 538). Use a proposal distribution  $x_1^* \sim N(x_1^* \mid x_1^{(\tau)}, 1.0)$  and  $x_2^* \sim N(x_2^* \mid x_2^{(\tau)}, 1.0)$  (simultaneously, not

alternately as in Gibbs sampling). Plot 50 samples with a line connecting consequtive samples '-' and rejected proposals with a cross 'x'.