

MDL exercises, fourth handout  
(due March 20th)

1. Let  $H(p) = -p \log p - (1-p) \log(1-p)$  denote the binary entropy of a Bernoulli[p] distribution when the probability of observing a zero is  $p$ . (The logarithm base is two.) Use Stirling's approximation  $\ln(n!) = (n + \frac{1}{2}) \ln n - n + \frac{1}{2} \ln 2\pi + O(1/n)$  to show that  $\log \binom{n}{\gamma n} = nH(\gamma) - \frac{1}{2} \log n + O(1)$ .
2. Suppose we observe an individual sequence  $X^n$  of 0s and 1s of length  $n$ , with  $n_1$  ones. Let  $\theta$  represent that the data are i.i.d. Bernoulli with probability of 1 given by  $\theta$ . Suppose first (a) that  $n_1/n = 1/3$ . For what value of  $\theta$  is  $P_\theta(x^n)$  maximized? Now suppose (b) that  $\theta = 1/3$ . For what value of  $n_1$  is  $P_\theta(x^n)$  maximized? Finally, suppose (c) again that  $\theta = 1/3$ . We sample  $n$  outcomes from  $P_\theta$ , and now we regard  $N_1$  as a random variable. For what values of  $m$  is  $P_\theta(N_1 = m)$  approximately maximized? (it is very hard to calculate this precisely, so it is sufficient to give two sets  $B_1$  and  $B_2$  (whose definition may depend on  $n$ ) such that for large  $n$ , for some  $m \in B_1$ , the probability is much larger than for all  $m \in B_2$ ).
3. Markov Chains.
  - a) Compute the maximum likelihood estimator  $\hat{\theta} = (p_{0 \rightarrow 1}, p_{1 \rightarrow 1})$  for a binary first order Markov chain.
  - b) Draw  $X_1, X_2, X_3$  from an order 1 Markov chain. Are  $X_1$  and  $X_3$  dependent? What if you know the value of  $X_2$ ?