

# 18.600 Recitation 12

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Partial solutions available at [math.mit.edu/~visheshj](http://math.mit.edu/~visheshj)

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**Solution Problem 1. (a)** Since  $X$  and  $Y$  are independent,  $H(X, Y) = H(X) + H(Y) = 2 \log_2(6)$ .

**(b)**  $H(Z) = (1/36) \log_2(36) + (2/36) \log_2(36/2) + \dots + (6/36) \log_2(36/6) + (5/36) \log_2(36/5) + \dots + (1/36) \log_2(36)$ .

**(c)**  $H_X(Z) = \log_2(6)$ .

**(d)**  $H(X, Z) = H(X) + H_X(Z) = 2 \log_2(6)$ . We could have deduced this directly from part (a) by noting that knowing  $(X, Z)$  is equivalent to knowing  $(X, Y)$ .

**Solution Problem 2. (a)** By Shannon's Noiseless Coding Theorem, Bob can encode his message using at least  $H(X)$  and at most  $H(X) + 1$  bits in expectation. In our case,  $H(X) = (1/16) \log_2(16) + (4/16) \log_2(16/4) + (6/16) \log_2(16/6) + (4/16) \log_2(16/4) + (1/16) \log_2(16) \approx 2.03$ .

**(b)** We can use a Huffman code. For instance,  $0 \mapsto 0000, 1 \mapsto 001, 2 \mapsto 1, 3 \mapsto 01, 4 \mapsto 0001$ .

**(c)** Let  $Y = (X_1, \dots, X_{100})$ . Then,  $H(Y) = 100H(X)$ . By Shannon's Noiseless Coding Theorem, Bob can encode his message using at least  $H(Y)$  and at most  $H(Y) + 1$  bits in expectation. Therefore, the expected number of bits *per symbol* is between  $H(X)$  and  $H(X) + (1/100)$ .

**Solution Problem 3. (a)** Note that

$$\mathbf{KL}(p||q) = \sum_{i=1}^n p_i \cdot (-\log_2(q_i/p_i)) \geq -\log_2\left(\sum_{i=1}^n p_i \cdot (q_i/p_i)\right) = -\log_2(1) = 0.$$

Here, we have used the convexity of the function  $x \mapsto -\log(x)$ . Equality holds if and only if equality holds in Jensen's inequality, which holds if and only if  $q_i/p_i$  does not depend on  $i$ , which is true if and only if  $p = q$ .

**(b)** We have

$$\mathbf{KL}(p||\text{Unif}) = \sum_{i=1}^n p_i \log_2(np_i) = -H(p) + \log_2(n).$$

Hence, by the previous part, we have  $\log_2(n) \geq H(p)$ , and equality holds if and only if  $p = \text{Unif}$ .

**Solution Problem 4.** Following the proof of the Chernoff bound, we have that for any  $\lambda > 0$ ,

$$\Pr[X \geq (p + \delta)n] = \Pr[\exp(\lambda X) \geq \exp(\lambda(p + \delta)n)] \leq \exp(-\lambda(p + \delta)n) \mathbb{E}[\exp(\lambda X)].$$

Moreover, we know that

$$\mathbb{E}[\exp(\lambda X)] = (pe^\lambda + (1 - p))^n.$$

Minimizing the quantity

$$\left( \frac{pe^\lambda + (1-p)}{e^{\lambda(p+\delta)}} \right)^n$$

over  $\lambda > 0$  gives us

$$e^\lambda = \frac{(1-p)(p+\delta)}{p(1-p-\delta)},$$

and substituting this into the previous expression shows that the minimum is

$$\left( \left( \frac{p}{p+\delta} \right)^{p+\delta} \left( \frac{1-p}{1-p-\delta} \right)^{1-p-\delta} \right)^n.$$

Finally, note that

$$\left( \frac{p}{p+\delta} \right)^{p+\delta} \left( \frac{1-p}{1-p-\delta} \right)^{1-p-\delta} = \exp(-\mathbf{KL}(\text{Ber}(p+\delta) \parallel \text{Ber}(p))).$$