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4. (30 points)

(a) Answer:

From (ii) and (iii), we get

$$r_{i} = H(Q_{\Delta_{i}}(Y_{i}))$$

$$= h(Y_{i}) - \log_{2} \Delta_{i}$$

$$\Rightarrow \Delta_{i} = 2^{h(Y_{i}) - r_{i}}$$

$$\rightarrow (i) \Rightarrow D_{i} = \frac{1}{12} 2^{2h(Y_{i})} 2^{-2r_{i}}.$$

$$(1)$$

Therefore, the optimization problem is a CMP to minimize

$$D(\mathbf{r}) \triangleq \sum_{i=1}^{k} D_i = \frac{1}{12} \sum_{i=1}^{k} 2^{2h(Y_i)} 2^{-2r_i}$$
 (2)

subject to $\sum_{i=1}^k r_i \leq R$. Since D_i is strictly decreasing in r_i , it follows that $\sum_{i=1}^k r_i = R$ for the optimal $\mathbf{r} = (r_1, \dots, r_k)$.

To use the Lagrange multiplier method, let

$$\begin{split} f(\mathbf{r}) &\triangleq D(\mathbf{r}) = \frac{1}{12} \sum_{i=1}^{k} 2^{2h(Y_i)} 2^{-2r_i}, \\ g(\mathbf{r}) &\triangleq \sum_{i=1}^{k} r_i - R, \\ J(\mathbf{r}, \lambda) &\triangleq f(\mathbf{r}) + \lambda g(\mathbf{r}). \end{split}$$

Then solving

$$\frac{\partial}{\partial r_j}J(\mathbf{r},\lambda) = \frac{1}{12}2^{2h(Y_i)}(-2\ln 2)2^{-2r_i} + \lambda = 0$$

yields

$$r_i = h(Y_i) + \underbrace{\log_2 \frac{\ln 2}{12\lambda}}_{\hat{j}}.$$

From the constraint $\sum_{j=1}^k r_j = R$, $\hat{\lambda}$ must satisfy

$$R = \sum_{j=1}^{k} h(Y_j) + k\hat{\lambda}$$
$$\Rightarrow \hat{\lambda} = \frac{R}{k} - \frac{1}{k} \sum_{j=1}^{k} h(Y_j).$$

Substituting into

$$r_i = h(Y_i) + \hat{\lambda}s$$

yields

$$r_i = h(Y_i) + \frac{R}{k} - \frac{1}{k} \sum_{j=1}^{k} h(Y_j).$$
 (3)

Then by (1) we get that $\forall i = 1, ..., k$,

$$\Delta_i = 2^{-\frac{R}{k} + \frac{1}{k} \sum_{j=1}^k h(Y_j)}.$$

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Hence, substituting (3) into (2) gives the the minimum total distortion as

$$D(\mathbf{r}) = \frac{1}{12} \sum_{i=1}^{k} 2^{2h(Y_i)} 2^{-2\left(\frac{R}{k} + h(Y_i) - \frac{1}{k} \sum_{j=1}^{k} h(Y_j)\right)}$$
$$= \frac{k}{12} 2^{\frac{2}{k} \sum_{j=1}^{k} h(Y_j)} 2^{-\left(\frac{R}{k}\right)}.$$

The above minimum distortion is the same as if k copies of a random variable with differential entropy $\frac{1}{k} \sum_{j=1}^k h(Y_j)$ were quantized using the uniform rate allocation $r_i = \frac{R}{k}$.

(b) **Answer:**

Since we have that $\forall i=1,\ldots,k,\ Y_i\sim N(\mu,\sigma_i^2)$, it follows that

$$\begin{split} h(Y_i) &= \frac{1}{2} \log_2(2\pi e \sigma_i^2) \\ \Rightarrow \frac{1}{k} \sum_{j=1}^k h(Y_j) &= \frac{1}{k} \sum_{j=1}^k \frac{1}{2} \log_2(2\pi e \sigma_i^2) \\ &= \frac{1}{2} \log_2(2\pi e) + \frac{1}{2} \log_2 \left(\left(\prod_{j=1}^k \sigma_j^2 \right)^{\frac{1}{k}} \right). \end{split}$$

Then by part (a),

$$D(\mathbf{r}) = \frac{k}{12} 2^{\frac{2}{k}} \sum_{j=1}^{k} h(Y_j) 2^{-\left(\frac{R}{k}\right)}$$
$$= \frac{k\pi e}{6} \left(\prod_{j=1}^{k} \sigma_j^2\right)^{\frac{1}{k}} 2^{-2\left(\frac{R}{k}\right)}.$$

(c) Answer:

Proof.