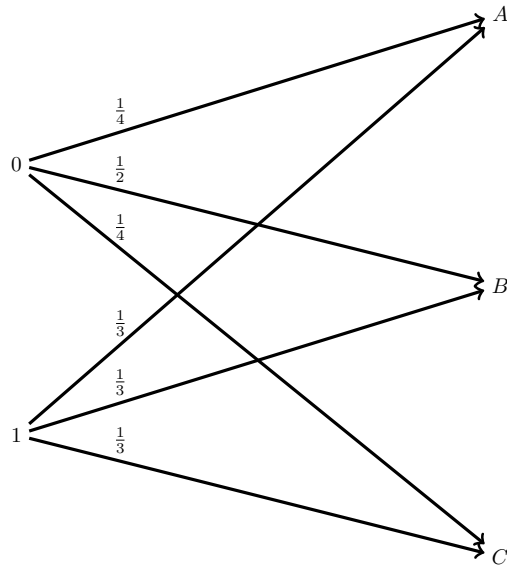


1. (2 points) Consider the binary input ternary output channel below. The inputs are either 0 or 1. The prior probability of the input being 0 is 0.4. The output is one of three symbols  $A, B, C$ .
- Find the decision rule  $\delta_{MPE}$  which minimizes the decision error probability.
  - Find the decision error probability when  $\delta_{MPE}$  is used.



2. (4 points) Consider the following binary hypothesis testing problem where the hypotheses are equally likely.

$$H_0 : Y \sim U \left[ -\sqrt{\frac{e^2\pi}{2}}, \sqrt{\frac{e^2\pi}{2}} \right]$$

$$H_1 : Y \sim \mathcal{N}(0, 1)$$

$U$  denotes the uniform distribution,  $\mathcal{N}$  denotes the Gaussian distribution and  $e$  is the base of the natural logarithm.

- Find the decision error probability of the rule which decides  $H_1$  is true if  $|Y| > \sqrt{\frac{e^2\pi}{2}}$  and decides  $H_0$  is true if  $|Y| \leq \sqrt{\frac{e^2\pi}{2}}$ . Express your answer in terms of the  $Q$  function.
  - Find the decision error probability of the optimal decision rule. Express your answer in terms of the  $Q$  function.
3. (2 points) Suppose observations  $Y_i, i = 1, 2, \dots, N$  are Poisson distributed with parameter  $\lambda$ . Assume that the  $Y_i$ 's are independent.
- Derive the ML estimator for  $\lambda$ .
  - Find the mean and variance of the ML estimate.
4. (2 points) Suppose we observe  $Y_i, i = 1, 2, \dots, M$  such that

$$Y_i \sim U[-\theta, \theta]$$

where  $Y_i$ 's are independent and  $\theta$  is unknown. Assume  $\theta > 0$ . Derive the ML estimator of  $\theta$ .

5. (2 points) Suppose observations  $X_i$  and  $Y_i$  ( $i = 1, \dots, N$ ) depend on an unknown parameter  $A$  as per the following distributions.

$$X_i \sim \mathcal{N}(A, \sigma^2), \quad i = 1, 2, \dots, N$$

$$Y_i \sim \mathcal{N}(A, 2\sigma^2), \quad i = 1, 2, \dots, N$$

Note that the variance of  $Y_i$  is twice the variance of  $X_i$ . Assume that  $X_i$  and  $X_j$  are independent for  $i \neq j$ . Assume that  $Y_i$  and  $Y_j$  are independent for  $i \neq j$ . Assume that  $X_i$  and  $Y_j$  are independent for all  $i, j$ . Assume  $\sigma^2$  is known.

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- (a) Derive the ML estimator for  $A$ .
- (b) Find the mean and variance of the ML estimate.
6. (2 points) Specify a method to generate a random variable with Rayleigh distribution which is a continuous random variable with probability distribution function given by

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 - e^{-\frac{x^2}{2\sigma^2}} & \text{otherwise} \end{cases}$$

where  $\sigma$  is a known parameter.

7. (2 points) Suppose a fair coin is repeatedly tossed. Specify a method to generate a random variable which corresponds to the number of tosses until a heads appears. Prove that the generated random variable has the correct distribution.

#### Useful Formulas

- Gaussian vector density  $p(\mathbf{x}) = \frac{1}{\sqrt{(2\pi)^n \det(\mathbf{C})}} \exp\left(-\frac{1}{2}(\mathbf{x} - \mathbf{m})^T \mathbf{C}^{-1}(\mathbf{x} - \mathbf{m})\right)$  where  $\mathbf{m} = E[\mathbf{X}]$ ,  $\mathbf{C} = E[(\mathbf{X} - \mathbf{m})(\mathbf{X} - \mathbf{m})^T]$
- Minimum probability of error rule

$$\delta_{\text{MPE}}(\mathbf{y}) = \arg \max_{1 \leq i \leq M} \pi_i p_i(\mathbf{y})$$

- The MAP decision rule is given by

$$\delta_{\text{MAP}}(\mathbf{y}) = \arg \max_{1 \leq i \leq M} P \left[ H_i \text{ is true} \middle| \mathbf{y} \right]$$

- The ML decision rule is given by

$$\delta_{\text{ML}}(\mathbf{y}) = \arg \max_{1 \leq i \leq M} p_i(\mathbf{y})$$