

Section 5

EE278: Introduction to Statistical Signal Processing (Fall 2020)

Monday, Oct 19, 2020 - 3 to 4 pm

1. Sufficient statistics

Let $V = v(\mathbf{Y})$ be a function of \mathbf{Y} for a binary hypothesis X with a discrete observation \mathbf{Y} . The following (for all sample values \mathbf{y}) are equivalent for $v(\mathbf{y})$ to be a sufficient statistic:

1. A function u exists such that $\Lambda(\mathbf{y}) = u(v(\mathbf{y}))$.
2. For any given positive a priori probabilities (i.e. p_X), the a posteriori probabilities satisfy

$$p_{X|\mathbf{Y}}(x | \mathbf{y}) = p_{X|\mathbf{Y},V}(x | \mathbf{y}, v(\mathbf{y})) = p_{X|V}(x | v(\mathbf{y}))$$

This means that $X - V - \mathbf{Y}$ is a Markov chain, i.e. X and \mathbf{Y} are conditionally independent given V .

3. The likelihood ratio of \mathbf{y} is the same as that of $v(\mathbf{y})$, i.e.

$$\Lambda(\mathbf{y}) = \frac{p_{V|X}(v(\mathbf{y}) | 1)}{p_{V|X}(v(\mathbf{y}) | 0)}$$

2. Whitening the Noise

1. Let $\mathbf{Z} = A\mathbf{W}$ where $\mathbf{W} \sim \mathcal{N}(0, I)$ is normalized IID Gaussian and A is non-singular. The observation random vector \mathbf{Y} is $\mathbf{a} + \mathbf{Z}$ given $X = \mathbf{a}$, and $\mathbf{b} + \mathbf{Z}$ given $X = \mathbf{b}$. Suppose the observed sample \mathbf{y} is transformed into $\mathbf{v} = A^{-1}\mathbf{y}$. Explain why \mathbf{v} is a sufficient statistic for this detection problem (and thus why MAP detection based on \mathbf{v} must yield the same decision as that based on \mathbf{y}).
2. Consider the detection problem where $\mathbf{V} = A^{-1}\mathbf{a} + \mathbf{W}$ given $X = \mathbf{a}$ and $\mathbf{V} = A^{-1}\mathbf{b} + \mathbf{W}$ given $X = \mathbf{b}$. Find $LLR(\mathbf{v})$ for a sample value \mathbf{v} of \mathbf{V} . Show that this is the same as the LLR for a sample value $\mathbf{y} = A\mathbf{v}$ of \mathbf{Y} .
3. Find $\Pr\{e | X = \mathbf{a}\}$ and $\Pr\{e | X = \mathbf{b}\}$ for the detection problem in (b). Show that you get the same answer by evaluating using the conditional densities from (a) directly.

Note: The methodology here is to transform the observed sample value to make the noise IID; this approach is often both useful and insightful.

References

- [1] R. G. Gallager. *Stochastic processes: theory for applications*. Cambridge University Press, 2013.