Statistical Inference and Multivariate Analysis (MA324)

Lecture 10

Point Estimation: Sufficiency, Factorization Theorem



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Jan-May 2023



Information and Sufficiency

- Aim is to estimate unknown parameter θ based on a realization of a RS using a suitable statistic or estimator.
- The RS $X=(X_1,X_2,\ldots,X_n)$ has all the "information" regarding unknown parameter θ . One should use a statistic that has same amount of "information" that the data have regarding θ .
- We can take T(X) = X. However, it is **not interesting** in most of the situations as one should take a summary of the data that capture all the "information".
- Therefore, in most of the cases, we will consider a function $T:\chi^n\to\mathbb{R}^m$, where m< n. In most of the times, the value of m is much smaller than that of n. Such summary or statistic is as good as the whole data and is called sufficient for θ .

Sufficient Statistic

- If a quantity vary with the change in another quantity, then there is some information in the first quantity regarding the second. On the other hand, if the first quantity do not change with the second quantity, then the first does not have any information regarding the second.
- Similarly, if the distribution of a statistic does not involve the unknown parameter θ, then the statistic does not have any information regarding θ. Motivated by this understanding, a sufficient statistic for θ can be defined as follows.

Def: [Sufficient Statistic] A statistic T = T(X) is called a sufficient statistic for unknown parameter θ if the conditional distribution of X given T = t does not include θ for all t in the support of T.

• Thus, given the value t of a sufficient statistic T, conditionally there is no information left in X regarding θ .

Sufficient Statistic

- In other words, X is trying to tell us a story regarding θ and any statistic is a gist of the story. If we have the gist T, a sufficient statistic, the original story is redundant as the gist has all the information that the original story has regarding θ .
- Note that X is a sufficient statistic. However, we are interested in a summary statistic in most of the situations.

Example 1: Let $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} Bernoulli(p), p \in (0, 1)$. Take $T = \sum_{i=1}^n X_i$. Show that T is a sufficient statistic for p.

- We can verify if a statistic is sufficient or not using the definition of sufficient statistic. That means that we first need to guess a correct statistic and then we can use the definition to show that it is actually a sufficient statistic for the unknown parameters.
- However, the next theorem gives necessary and sufficient conditions, which can be used to find a sufficient statistic. Therefore, the next theorem is very useful.

Theorem (Neyman-Fisher Factorization Theorem)

Let X_1, \ldots, X_n be RS with JPMF or JPDF $f_{\boldsymbol{X}}(\boldsymbol{x}, \boldsymbol{\theta}), \boldsymbol{\theta} \in \Theta$. Then $T = T(X_1, \ldots, X_n)$ is sufficient for $\boldsymbol{\theta}$ if and only if

$$f_{\mathbf{X}}(\mathbf{x}, \, \boldsymbol{\theta}) = h(\mathbf{x})g_{\boldsymbol{\theta}}\left(\mathbf{T}(\mathbf{x})\right),$$

where h(x) does not involve θ , $g_{\theta}(\cdot)$ depends on x only through T(x).

Example 2: Let $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} Poi(\lambda), \lambda > 0$. Find a sufficient statistic for λ .

Example 3: Let $X_1, X_2, \ldots, X_n \stackrel{i.i.d.}{\sim} N(\mu, \sigma^2), \mu \in \mathbb{R}$ and $\sigma > 0$. Find a sufficient statistic for (μ, σ^2)