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3.5. Axiomatization of Information Measurement

Consequently the student who is devoid of talent will derive no more profit from this work than barren soil from a treatise on agriculture, Quintilian (35-95)

Information entropy (quantity of information) has an axiomatic description. Namely, there is a set of axioms, which characterize some natural properties of this information measure, and these axioms uniquely determine Shannon's entropy.

There are different systems of axioms used for characterization of the information entropy. We follow here the approach of Shannon (1948).

Building his axiomatic system based on natural properties of information measure, Shannon starts with the problem of information transmission. He analyses how to define a quantity that will measure, in some sense, how much information is "produced" by a discrete information source, or as Shannon explains, at what rate information is produced. With such a goal, the problem is specified by taking into account only information that tells the receiver whether some event has happened. To estimate this information, it is assumed that the receiver regards a set $\{E_1, E_2, ..., E_n\}$ of possible events with known probabilities of occurrence $p_1, p_2, ..., p_n$. Then it is possible to assume that the measure for the information entropy $H(p_1, p_2, ..., p_n)$ should have the following properties:

Axiom A1 (*Continuity*). The function $H(p_1, p_2, ..., p_n)$ is continuous in all arguments p_i .

Axiom A2 (Monotonicity). If all p_i are equal $(p_i = 1/n)$, then H(1/n, 1/n, ..., 1/n) must increase monotonically with n,

Axiom A3 (Branching property). If a choice is broken down into two successive choices, the original function $H(p_1, p_2, ..., p_n)$ should be the weighted sum of the two new values of this function and the weights are equal to the probabilities of branching.

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\pagestyle { headings }
\author{Karlis Kreilis}
\del{date} \{ March 2019 \}
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\newpage
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    \textit {\scriptsize { Statistical Information Theory }}
\end{center}
\noindent \scalebox {.9}[1.0]
{\textbf{3.5. Axiomatization of Information Measurement}}
\begin{flushright}
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\setminus setcounter { page } {3}
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