

## Discrete Variable's Measurement Using Non-numerical, Inexact, and Incomplete Information on Probabilities' Distribution

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When measuring a discrete variable  $x$ , whose possible values constitute a finite set  $X = \{x_1, \dots, x_n\}$ , one uses, as a rule, the stochastic model, suggesting that the observed results obtained in the measurement of the variable are realizations of a random variable  $\tilde{x}$ , this variable being specified by a probabilities' distribution  $p = (p_1, \dots, p_n)$ , where  $p_i$  is a probability that the corresponding value  $x_i$  will appear. Within this model's framework one usually chooses a mathematical expectation  $\mu = E\tilde{x} = x_1 \cdot p_1 + \dots + x_n \cdot p_n$  of the random variable  $\tilde{x}$  as an estimation of the discrete variable under measuring.

Unfortunately, in a real measurement practice, we usually have to deal with an *uncertainty* of probabilities' vector  $p = (p_1, \dots, p_n)$  determination, this uncertainty being manifested in the fact that the vector is represented within an accuracy of a certain set  $P(n; I)$  of all admissible (from the point of view of an additional information  $I$ ) probabilities' vectors. In so doing, the information  $I$ , which determines the set  $P(n; I)$ , is often *non-numerical, inexact, and incomplete* one. For modeling of the uncertainty of a probabilities' vector specification we propose to use the Bayesian strategy of "*randomization of uncertainty*". In the case considered, the uncertainty of the choice of a vector  $p = (p_1, \dots, p_n)$  from the set  $P(n; I)$  is modeled by a random vector  $\tilde{p} = (\tilde{p}_1, \dots, \tilde{p}_n)$ , which takes values from the set  $P(n; I)$ . The obtained *randomized (random) probabilities*  $\tilde{p}_1, \dots, \tilde{p}_n$  can be substituted into the expression for the expectation  $\mu$ , what will give the *randomized mathematical expectation*  $\tilde{\mu} = x_1 \cdot \tilde{p}_1 + \dots + x_n \cdot \tilde{p}_n$ , this randomized expectation being used as a model for the uncertainty of the discrete variable's estimation.

The report gives explicit formulae for mathematical expectation  $\bar{\mu}(I)$  and variance  $\bar{\sigma}^2(I)$  of the randomized expectation  $\tilde{\mu}$ , these formulae being used for a *numerical image*  $(\bar{\mu}(I), \bar{\sigma}^2(I))$  of *non-numerical, inexact and incomplete information*  $I$ , which is possessed by an investigator, who is carrying out a measurement of a discrete variable.