

Data D can be represented as an attribute-based matrix (x_{ij}) , $i = 1, \dots, m$; $j = 1, \dots, n$; where x_{ij} is the numerical value of j attribute on i -th object. Attributes may be qualitative and quantitative. The fact that numerical values of attributes exist means that there are n measurement procedures, which produce them. Let us denote these procedures as $x_j(a)$, $j = 1, \dots, n$, where a is an empirical entity. Then, $x_{ij} = x_j(a_i)$.

Attribute-value methods such as neural networks deal with this type of data. These methods are limited by the assumption that data types are stronger than interval and log-interval data types, which is not always true. For definitions of these data types, see Section 4.9.2.

Let us determine an empirical axiomatic system for attribute-based matrices. At first, a set of empirical predicates V_i for each attribute x_i needs to be defined. There are two cases:

1. The measurement procedure x_i is well known and an empirical system is known from measurement theory. Therefore, the set of all empirical predicates contains predicates given in V_i $i=1, \dots, n$.

2. An empirical system of the measurement procedure x_i is not completely defined. In this case, we have a measurement procedure, but we do not have an empirical system. The measurement procedure in the second case is called a measurer. The examples of measurers are psychological tests, stock market indicators, questionnaires, and physical measurers used in non-physical areas.

Let us define a set of empirical predicates V_i for measurer procedure x_i . For any numerical relation $R(y_1, \dots, y_k)$ in Re^k (Re - the set of all real numbers), we can define the following empirical relation on A^k .

$$P^R(a_1, \dots, a_k) \Leftrightarrow R(x_i(a_1), \dots, x_i(a_k)).$$

The measurer x_i obviously has an empirical interpretation, but relation P^R may not. We need to find such relations R that have empirical interpretations, i.e., relation $R(x_i(a_1), \dots, x_i(a_k))$ is interpretable in the terms of domain theory.

Suppose that $\{R_1, \dots, R_k\}$ is a set of the most common numerical relations and some (relations $P_j^{R_1}, \dots, P_j^{R_k}$) have an empirical interpretation. This set of relations is not empty, because at least the relation P_j^- (equivalence) has an empirical interpretation:

$$P_j^-(a_1, a_2) \Leftrightarrow x_j(a_1) = x_j(a_2).$$

In measurement theory, there are many sets of axioms based on just ordering and equivalence relations. Nevertheless, these sets of axioms establish strong data types. A strong data type is a result of interaction of the quantities with individual weak data types such as ordering and equivalence. For instance, having one weak order relation (for attribute y) and n equivalence relations

$$\{\leq_y, =_{x_1}, \dots, =_{x_i}, \dots, =_{x_n}\}$$

for attributes x_1, \dots, x_n , we can construct a complex relation between y and x_1, \dots, x_n given by

$$G(y, x_1, \dots, x_n) \Leftrightarrow y = f(x_1, \dots, x_n),$$

where $f(x_1, \dots, x_n)$ is a polynomial [Krantz et al, 1971].

This is a very strong result. To construct a polynomial we need the sum operation, but this operation is not defined for x_1, \dots, x_n . However, relation G is equivalent to polynomial f if a certain set of axioms expressed in terms of order relation ($<_y$) presented above for y , and equality relations ($=$) are true for x_i .

This fundamental result serves as a critical justification for using ordering relations as a base for generating relational hypotheses in financial applications (Chapter 5). Ordering relations usually are empirically interpretable in finance. Multivariate and pair comparisons [Torgerson, 1952, 1958, Shmerling D.S. 1978]. Consider set of objects $A = \{a_1, \dots, a_m\}$ and set of all tuples A^k of k objects from A . A group of n experts are asked to order objects in all tuples $\langle a_1, a_2, \dots, a_k \rangle$ from A^k in accordance with some preference relation. Let a_i^{tsq} be an object i from tuple $\langle a_1, a_2, \dots, a_k \rangle$, where t is an entity to be evaluated, s is an expert and q is a preference rank given by an expert s to the entity t , $i = 1, \dots, m$; $s = 1, \dots, n$; $t = 1, \dots, C_m^k$; $q = 1, \dots, k$. The set of all ordered tuples is denoted by

$$R = \{ \langle a_{i1}^{ts_1}, a_{i2}^{ts_2}, \dots, a_{ik}^{ts_k} \rangle \}.$$

The typical goal of pair and multivariate comparison methods is to order all tuples. Known methods are based on some a priori assumptions [Torgerson, 1952, 1958; Shmerling D.S. 1978] which determine the areas of applicability. Let us define for every expert s the preference relation

$$P_s(a_{i1}^{ts_1}, a_{i2}^{ts_2}) \Leftrightarrow i1 < i2.$$

Also, define the two equivalence relations \sim , \sim_t and equivalence relation $=$ by

$$a_{i1}^{t1s1} \sim a_{i2}^{t2s2} \Leftrightarrow i1 = i2,$$

$$a_{i1}^{t1s1} \sim_t a_{i2}^{t2s2} \Leftrightarrow t1 = t2, \text{ and}$$

$$a_{i1}^{ts_1} = a_{i2}^{ts_2} \Leftrightarrow \text{objects } a_{i1}^{ts_1}, a_{i2}^{ts_2} \text{ are the same.}$$

Therefore, we obtain a set of empirical predicates

$$V = \{=, \sim, \sim_t, P_1, \dots, P_n\}$$

and, thus, a data type represented with the set of empirical predicates V .