

# Chapter 11

## MV-Partitions and MV-Powers

**C. Drossos**

*University of Patras, Greece*

**P. L. Theodoropoulos**

*Educational Counselor, Greece*

### ABSTRACT

*In this chapter, the authors generalize the Boolean partition to semisimple MV-algebras. MV-partitions together with a notion of refinement is tantamount a construction of an MV-power, analogous to Boolean power construction (Mansfield, 1971). Using this new notion we introduce the corresponding theory of MV-powers.*

### 1. INTRODUCTION

In this introduction we should like to touch on some points that will help the readers to comprehend the contents of the Chapter.

#### 1.1 Interpretations and the Change of Point of View

In modern mathematics we usually use Set Theory to express mathematical notions. In Set theory, as such there, is no uncertainty involved. However we usually have the habit to interpret some of the set-theoretic notions, in such a way that we model uncertainty. For example the func-

tion  $\varphi : X \rightarrow [0,1]$  is a well-defined entity in Set Theory, and no uncertainty is connected with such a function. However we may interpret this as a generalized indicator or a membership function of a fuzzy set. This change of point of view leads to Fuzzy Set theory. Similarly a real number  $r \in \mathbb{R}$ , is interpreted in fuzzy set theory as a fuzzy real number  $\tilde{r} : \mathbb{R} \rightarrow [0,1]$ . Related to the change of point of view the following Dieudonne's saying is instructive: "What changes in Mathematics, as in all other Sciences, is the point of view from which results already acquired, are assessed."

The change of point of view essentially is connected with *interpretations* from one model to another. Think, for example, the interpretation of Euclidean terms in Riemannian geometry. To give a technical development of the intuition

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of interpretations it is a hard work, see e.g. (W. Hodges, 1993, Ch. 5).

Similarly, a functor, e.g. going from the category of Topological Spaces to the Category of Groups, can be construed as a *change of point of view*, from a topological-geometrical to algebraic.

### 1.2 Generalized Elements and Generalized Properties

Since functions play a very important role in changing a point of view from a classical set theoretic model to a non-classical one, we should like to comment on the two basic interpretations for the concept of function:

- **Variable or Generalized Elements:** Instead of classical elements  $(x \in X) \equiv (x : 1 \rightarrow X)$  we consider variable ones  $x : T \rightarrow X$ .  $T$  is considered as the domain of variation or the stage of definition of the variable element.
- **Generalized Properties:** Every indicator function  $I_A : X \rightarrow 2 := \{0, 1\}$  describes a crisp property  $p := \{x \in X : I_A = 1\}$ . We would like to generalize this to have instead of 2, any set  $V := \{v_1, v_2, \dots, v_n\}$  or  $V = [0, 1]$  and in general any set  $V$ .

Thus a fuzzy set  $\varphi : X \rightarrow [0, 1]$  from this point of view is a generalized property. Any generalized property  $p : X \rightarrow V$  induces on  $X$  a partition,

$$(\varphi^{-1}(t))_{t \in [0,1]}$$

In this way partitions and generalized properties are strongly interconnected.

One general approach to this and similar problems is through Category Theory. We know that there is a “reversing the arrows duality” between subsets and partitions see (Lawvere & Rosebruch, 2003 and Ellerman, 2010). In particular we are interesting in the duality of “generalized elements” and “generalized properties” (Lawvere & Rosebruch, 2003).

We would like to present this reversing the arrows duality in some characteristic examples.

Let first consider a function  $f : A \rightarrow B$  as a diagram in the category of sets, using as an index the graph  $\bullet \rightarrow \bullet$ . Then the graph  $G_f$  is the limit of the diagram  $f : A \rightarrow B$  as depicted in Figure 1.

In this way we may have a notion of a graph of an arrow in any category with finite limits.

To reveal the connection of Boolean powers with the reversing the arrows duality we have to calculate explicitly the reverse of the function  $f : A \rightarrow B$ .

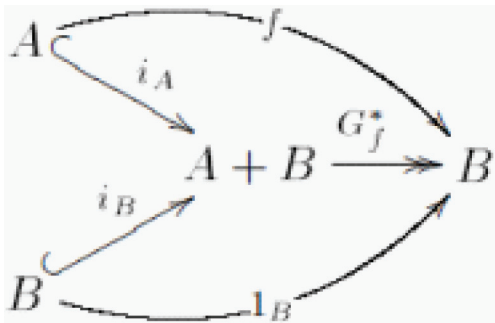
On the domain  $A$  we define the equivalence relation:

$$a_1 \approx_f a_2 :\Leftrightarrow f(a_1) = f(a_2), \quad a_1, a_2 \in A$$

Let now,

$$A_b := \{a \in A \mid f(a) = b\} \quad b \in B$$

Figure 1. Limit of diagram  $f : A \rightarrow B$



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