# The description of extreme 2-monotone measures 

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#### Abstract

The paper is devoted to the description of extreme points in the set of 2 -monotone measures. We describe them using lattices on which an extreme 2 -monotone measure is additive. We also propose the way of generation extreme monotone measures based on the aggregation of extreme measures with the help of multilinear extension. We describe also the class of extreme 2 -monotone measures that are additive on the filter on which a 2 monotone measure has positive values.


Keywords. 2-monotone measures, extreme points, additivity on lattices, filters, partially ordered sets, multilinear extension.

## 1 Introduction

2-monotone measures play an important role in the theory of imprecise probabilities [17], because for imprecise probabilities represented by 2 -monotone measures it is possible to find analytical solutions for many problems and, therefore, such models are more attractive in a computational point of view. Meanwhile, some unsolved problems concerning 2-monotone measures can be solved [5] if we know the structure of extreme points of the set of all 2-monotone measures. It is worth to mention that finding description of extreme points of a convex set is usually a hard problem. This problem is solved for the set of all monotone measures [13,15], $p$-symmetrical measures $[7,8]$, but for some convex families, e.g. $k$-additive measures [7], is far from the final solution
The aim of this paper is to make one step forward in this direction, providing some general necessary and sufficient conditions that a 2 -monotone measure is an extreme point and giving descriptions of some special families of them.
The paper has the following structure. We remind first some results concerning monotone measures and criteria of 2 -monotonicity. After that we provide general necessary and sufficient conditions that a 2 -monotone measure is an extreme point through lattices on which it is additive. After that we remind the multilinear extension of monotone measures and using it we define the composition of monotone measures. We show that the composition of extreme 2-monotone measures is an extreme 2-monotone measure again. The paper is ended

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by describing a special class of 2-monotone measures which are additive on the filter of sets on which a 2 monotone measure has positive values.

## 2 Monotone measures

Let $X$ be a finite set and let $\mu: 2^{X} \rightarrow[0,1]$ be a set function on the powerset $2^{X}$. Then $\mu$ is called a monotone measure [9] if the following conditions hold:

1) $\mu(\varnothing)=0$ and $\mu(X)=1$;
2) $A \subseteq B$ for $A, B \in 2^{X}$ implies $\mu(A) \leq \mu(B)$.

Let us denote the set of all monotone measures on $2^{X}$ by $M_{\text {mon }}(X)$ or briefly $M_{\text {mon }}$ if the set $X$ is clearly defined from the context. For monotone measures $\mu_{1}, \mu_{2} \in M_{\text {mon }}$ we define their convex sum as $\mu(A)=a \mu_{1}(A)+$ $(1-a) \mu_{2}(A)$, where $a \in[0,1]$ and $A \in 2^{X}$. Clearly, $\mu \in M_{\text {mon }}$, i.e. the set $M_{\text {mon }}$ is convex and it is possible to show $[13,15]$ that extreme points of $M_{\text {mon }}$ are $\{0,1\}$ valued monotone measures, i.e. monotone measures with values in $\{0,1\}$. Let the algebra $2^{X}$ be considered as a partially ordered set w.r.t. inclusion of sets. By definition, a filter $\mathbf{f}$ in $2^{X}$ is a nonempty subset of $2^{X}$ such that $A \in \mathbf{f}, A \subseteq B$ implies $B \in \mathbf{f}$. Any filter can be uniquely defined by the set of its minimal elements $\left\{A_{1}, \ldots, A_{m}\right\}$. This fact is denoted by $\mathbf{f}=\left\langle A_{1}, \ldots, A_{m}\right\rangle$. The connection between filters of algebra $2^{X}$ and $\{0,1\}$ valued monotone measures is shown in the following lemma [13].
Lemma 1. Any $\{0,1\}$-valued monotone measure $\eta$ defines a filter $\mathbf{f}=\left\{A \in 2^{X} \mid \eta(A)>0\right\}$ such that $\varnothing \notin \mathbf{f}$. Conversely, any filter $\mathbf{f}$ with $\varnothing \notin \mathbf{f}$ defines a $\{0,1\}$ valued monotone measure $\eta$ by

$$
\eta(A)= \begin{cases}1, & A \in \mathbf{f},  \tag{1}\\ 0, & A \notin \mathbf{f} .\end{cases}
$$

In the sequel we denote a $\{0,1\}$-valued measure as $\eta_{\mathrm{f}}$ if it corresponds to a filter $\mathbf{f}$.
Remark 1. Clearly, a set $\mathbf{f}(t)=\left\{A \in 2^{X} \mid \mu(A)>t\right\}$ for any given $\mu \in M_{\text {mon }}$ and $t \in[0,1)$ is a filter in algebra
$2^{X}$, moreover, $\mu(A)=\int_{0}^{1} \eta_{\mathbf{f}(t)}(A) d t$ and if $\left\{t_{1}, t_{2}, \ldots, t_{k}\right\}$ is the set of all values of $\mu$ and $0=t_{1}<t_{2}<\ldots<t_{k}=1$, then $\mu=\sum_{i=1}^{k-1}\left(t_{i+1}-t_{i}\right) \eta_{\mathbf{f}\left(t_{i}\right)}$.

## 3 2-monotone measures

A monotone measure $\mu$ is called 2-monotone [9] if the following inequality

$$
\begin{equation*}
\mu(A)+\mu(B) \leq \mu(A \cap B)+\mu(A \cup B) \tag{2}
\end{equation*}
$$

is fulfilled for any $A, B \in 2^{X}$. We denote the set of all 2monotone measures on the algebra $2^{X}$ by $M_{2-\text { mon }}(X)$. The condition (2) can be simplified [3]. It is sufficient to check inequalities of the following type:

$$
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right) \leq \mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right),(3)
$$

for all $A \in 2^{X}$ and $x_{i}, x_{j} \in 2^{X}$ such that $|X \backslash A| \geq 2$, $x_{i}, x_{j} \in X \backslash A$ and $x_{i} \neq x_{j}$.
In the next we can also consider nonnegative set functions $\mu$ on $2^{X}$ with $\mu(\varnothing)=0$. Such set functions are called 2 -monotone if they are monotone and inequalities (2) or equivalently inequalities (3) are fulfilled. The next proposition shows that the monotonicity of $\mu$ is not necessary to check.
Proposition 1. Let $\mu$ be a nonnegative set function on $2^{X}$ with $\mu(\varnothing)=0$. Then it is 2 -monotone iff inequalities (3) are fulfilled for all $A \in 2^{X}$ and $x_{i}, x_{j} \in 2^{X}$ such that $|X \backslash A| \geq 2, x_{i}, x_{j} \in X \backslash A$ and $x_{i} \neq x_{j}$.
Let us consider how Proposition 1 can be strengthened if we know that the sets on which a nonnegative set function is positive, form a filter. In this case we say that $\mu$ is 2-monotone on the filter $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$, if inequalities (3) are fulfilled, when $A \cup\left\{x_{i}\right\} \in \mathbf{f}$ and $A \cup\left\{x_{j}\right\} \in \mathbf{f}$. In addition, we say that a set function $\mu$, which is 2-monotone on the filter $\mathbf{f}$, is also 2-monotone on its borders if inequalities (3) are fulfilled if at least $A \cup\left\{x_{i}\right\} \in \mathbf{f}$ or $A \cup\left\{x_{j}\right\} \in \mathbf{f}$. Next proposition is the direct consequence of Proposition 1 .
Proposition 2. Given a nonnegative set function $\mu$ such that $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$ is a filter. Then $\mu$ is 2 monotone iff it is 2-monotone on the filter $\mathbf{f}$ and its borders.
In some cases the 2-monotonicity on the filter can imply the 2 -monotonicity on its borders. The description of such a case is given in the following proposition.
Proposition 3. Let a nonnegative set function be 2monotone on the filter $\mathbf{f} \supseteq\left\{A \in 2^{X} \mid \mu(A)>0\right\}$ and let $\mathfrak{C}=\left\{C_{1}, \ldots, C_{m}\right\}$ be the set of its minimal elements. Then $\mu$ is 2-monotone on borders of $\mathbf{f}$, if for every $C_{k} \in \mathfrak{C}$
and every $x_{i} \notin C_{k}$ there exists a $C_{l} \in \mathfrak{C}^{*}$, such that $\left\{x_{i}\right\}=C_{l} \backslash C_{k}$.

## 4 Additivity properties of 2-monotone measures on lattices

We denote by $M_{p r}(X)$ the set of all probability measures on the algebra $2^{X}$. Let $\mu \in M_{2-\text { mon }}$, then the core of $\mu$ is the set of probability measures defined by $\operatorname{core}(\mu)=\left\{P \in M_{p r} \mid P \geq \mu\right\}$. It is well known [16] that $\operatorname{core}(\mu)$ is a nonempty convex set for any $\mu \in M_{2-m o n}$ and its extreme points are probability measures $P_{\gamma}$, where $\gamma:\{1,2, \ldots, n\} \rightarrow\{1,2, \ldots, n\}$ is a permutation of the set $\{1,2, \ldots, n\}$ and any $P_{\gamma}$ is constructed with the help of the chain of sets $B_{1}=\left\{x_{\gamma(1)}\right\}, B_{2}=\left\{x_{\gamma(1)}, x_{\gamma(2)}\right\}, \ldots$, $B_{n}=\left\{x_{\gamma(1)}, \ldots, x_{\gamma(n)}\right\} \quad$ by the rule: $\quad P_{\gamma}\left(B_{i}\right)=\mu\left(B_{i}\right)$, $i=1, \ldots, n$. Let us remind the result from [2,4], that can be also found in [10].
Proposition 4. Let $\mu \in M_{2-\text { mon }}$, then the system of sets $\mathcal{L}_{\gamma}(\mu)=\left\{A \in 2^{X} \mid \mu(A)=P_{\gamma}(A)\right\} \quad$ is a lattice w.r.t. operations $\cap$ and $\cup$, i.e. $A, B \in \mathcal{L}_{\gamma}(\mu)$ implies $A \cap B, A \cup B \in \mathcal{L}_{\gamma}(\mu)$ and it is a maximal lattice, on which $\mu$ is additive.
Remark 2. Additivity of $\mu$ on $\mathcal{L}_{\gamma}(\mu)$ means that if $A, B \in \mathcal{L}_{\gamma}(\mu)$, then

$$
\mu(A)+\mu(B)=\mu(A \cap B)+\mu(A \cup B)
$$

As we will see in the next such maximal lattices play an important role for the extreme points description of 2monotone measures. Therefore, we also present here some results showing the connections between such lattices and partially ordered sets.
Let us assume that a maximal lattice $\mathfrak{L}$, on which a 2 monotone measure $\mu$ is additive, contains maximal chains described by a set of permutations $\Gamma=\left\{\gamma_{i}\right\}$. We put into correspondence the linear order $\rho_{\gamma}$ on $X$ to each permutation $\gamma \in \Gamma$ in a way that $x_{i} \rho_{\gamma} x_{j}$ if $\gamma(i) \leq \gamma(j)$. Then the following theorem is valid.
Theorem 1. Let $\mathfrak{L}$ be a maximal lattice, on which a 2monotone measure $\mu$ is additive, and let the maximal chains in $\mathfrak{L}$ be described by a a set of permutations $\Gamma=\left\{\gamma_{i}\right\}$. Consider the partial order $\rho_{\Gamma}=\bigcap_{\gamma \in \Gamma} \rho_{\gamma}{ }^{l}$, where linear orders $\rho_{\gamma}$ are defined as above. Then $\left\{\rho_{\gamma}\right\}_{\gamma \in \Gamma}$ is the set of all linear orders satisfying $\rho_{\gamma} \supseteq \rho_{\Gamma}$.

[^0]Next result can be considered as a corollary of more general result that can be found in [10].
Theorem 2. Let $\rho$ be a partial order on $X$ and let $\left\{\rho_{\gamma}\right\}_{\gamma \in \Gamma}$ be the set of all linear extensions $\rho$, i.e. each $\rho_{\gamma}$ is a linear order and $\rho_{\gamma} \supseteq \rho$. Then any $\rho_{\gamma}, \gamma \in \Gamma$, induces $a$ chain of sets $B_{0}=\varnothing, \quad B_{1}=\left\{x_{\gamma(1)}\right\}$, $B_{2}=\left\{x_{\gamma(1)}, x_{\gamma(2)}\right\}, \ldots, \quad B_{n}=\left\{x_{\gamma(1)}, x_{\gamma(2)}, \ldots, x_{\gamma(n)}\right\}$ and the union of all such chains is a lattice of sets w.r.t. union and intersection.

## 5 The description of extreme 2-monotone measures through lattices

Proposition 5. Let us consider the set of all maximal lattices $\mathfrak{L}_{\gamma}(\mu)$, on which a 2-monotone measure $\mu$ is additive, and let $\mathbf{f}_{\mu}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$. Then $\mu$ is not an extreme point in $M_{2-m o n}$ iff there exists a 2-monotone measure $v(v \neq \mu)$ such that

1) $\mathfrak{L}_{\gamma}(\mu) \subseteq \mathfrak{L}_{\gamma}(\nu)$ for any permutation $\gamma$;
2) $\mathbf{f}_{v} \subseteq \mathbf{f}_{\mu}$.

Corollary 1. Let $\mu$ be an extreme point in $M_{2-m o n}$. Then the filter $\mathbf{f}_{\mu}$ and the system of lattices $L_{\gamma}(\mu)$ define $\mu$ uniquely.

## 6 Multilinear extension and composition of monotone measures

In this section we will use the notion of pseudo-Boolean functions [12]. Any pseudo-Boolean function is a mapping $\varphi:\{0,1\}^{n} \rightarrow \mathbb{R}$. For our purpose, it is sufficient to consider pseudo-Boolean functions taking their values in $[0,1]$, i.e. we assume that $\varphi:\{0,1\}^{n} \rightarrow[0,1]$. It is easy to see that there is a one-to-one correspondence between pseudo-Boolean functions and set functions. For this purpose, we consider set functions defined on the algebra $2^{Z}$, where $Z=\{1, \ldots, n\}$, and consider vectors $\mathbf{1}_{A}=\left(x_{1}, \ldots, x_{n}\right)$, where $A \in 2^{Z}$ and $x_{i}=1$ if $i \in A$ and $x_{i}=0$ otherwise. Then obviously $\mu(A)=\varphi\left(\mathbf{1}_{A}\right)$, where $A \in 2^{Z}$ is a set function on $2^{Z}$. If we consider the class of monotone pseudo-Boolean functions $\varphi:\{0,1\}^{n} \rightarrow[0,1]$ with $\varphi(\mathbf{0})=0$ and $\varphi(\mathbf{1})=1$, where $\mathbf{0}=(0, \ldots, 0)$ and $\mathbf{1}=(1, \ldots, 1)$, then it corresponds to the class of monotone measures on $2^{Z}$.
Any pseudo-Boolean function can be uniquely represented as a multilinear polynomial [14] as

$$
\begin{equation*}
\varphi(\mathbf{x})=\sum_{A \in 2^{2}} m(A) \prod_{i \in A} x_{i}, \tag{4}
\end{equation*}
$$

where $m$ is the Möbius transform $m$ of the set function $\mu(A)=\varphi\left(\mathbf{1}_{A}\right)$, defined by

$$
m(A)=\sum_{B \subseteq A}(-1)^{|A| B \mid} \mu(B) .
$$

We see that there is a one-to-one correspondence between multilinear polynomials and pseudo-Boolean
functions. In addition, we can assume that the vector $\mathbf{x}$ in formula (4) can take values in $[0,1]^{n}$. In this case, the function $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ is called [14] the multilinear extension of $\varphi$.
The next proposition [3] shows how to check monotonicity and 2-monotonicity of a set function using its multilinear extension.
Proposition 6. Let $\mu: 2^{Z} \rightarrow[0,1]$ and let $\varphi$ be its corresponding pseudo-Boolean function. Then $\mu$ is a monotone measure iff the multilinear extension $\tilde{\varphi}$ of $\varphi$ has the following properties:

1) $\tilde{\varphi}(\mathbf{0})=0$ and $\tilde{\varphi}(\mathbf{1})=1$;
2) $\frac{\partial \tilde{\varphi}(\mathbf{x})}{\partial x_{i}} \geq 0$ for any $x_{i}$ and at any point $\mathbf{x} \in[0,1]^{n}$.

In addition, $\mu$ is 2-monotone iff
3) $\frac{\partial^{2} \tilde{\varphi}(\mathbf{x})}{\partial x_{i} \partial x_{j}} \geq 0$ for any $x_{i}, x_{j}$ and at any point $\mathbf{x} \in[0,1]^{n}$.

Proposition 6 shows that the multilinear extension of a monotone measure is an aggregation function. Let us remind that, by definition [11], an aggregation function $\tilde{\varphi}$ is a mapping $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ such that

1) $\tilde{\varphi}(\mathbf{0})=0$ and $\tilde{\varphi}(\mathbf{1})=1$;
2) $\tilde{\varphi}(\mathbf{x}) \leq \tilde{\varphi}(\mathbf{y})$ for $\mathbf{x}, \mathbf{y} \in[0,1]^{n}$ if $\mathbf{x} \leq \mathbf{y} \quad(\mathbf{x} \leq \mathbf{y}$ means for $\mathbf{x}=\left(x_{1}, \ldots, x_{n}\right)$ and $\mathbf{y}=\left(y_{1}, \ldots, y_{n}\right)$ that $x_{i} \leq y_{i}$, $i=1, \ldots, n)$.
We can generate monotone measures using aggregation functions as follows. Let $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ be an aggregation function and $X_{1}, \ldots, X_{n}$ be mutually disjoint finite nonempty sets and let $\mu_{i}, i=1, . ., n$, be monotone measures on $2^{X_{i}}$. Then a set function $\mu$ on $2^{X}$, where $X=X_{1} \cup \ldots \cup X_{n}$, defined by

$$
\begin{equation*}
\mu(A)=\tilde{\varphi}\left(\mu_{1}\left(A \cap X_{1}\right), \ldots, \mu_{n}\left(A \cap X_{n}\right)\right), A \in 2^{X} \tag{5}
\end{equation*}
$$

is also a monotone measure. For the measure $\mu$, defined by formula (5), we will use the notation $\mu=\tilde{\varphi} \circ \boldsymbol{\mu}$, where $\boldsymbol{\mu}=\left(\mu_{1}, \ldots, \mu_{n}\right)$.
In this section, we will use multilinear polynomials as aggregation functions. It can be shown [3] that if $\varphi$ is a multilinear extension of a 2 -monotone measure and $\mu_{i}$, $i=1, \ldots, n$, be 2 -monotone measures on $2^{x_{i}}$, then $\mu=\varphi \circ \mu$ is also a 2-monotone measure.
Obviously, we can introduce the same representation like (5) for pseudo-Boolean functions. Let $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ be an aggregation function and let $\mu_{i}\left(\mathbf{x}^{(i)}\right), i=1, \ldots, n$, be pseudo-Boolean functions. Then the aggregation of these functions is defined as
$\mu(\mathbf{x})=\tilde{\varphi}\left(\mu_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \mu_{n}\left(\mathbf{x}^{(n)}\right)\right)$, where $\mathbf{x}=\left(\mathbf{x}^{(1)}, \ldots, \mathbf{x}^{(n)}\right)$.
Proposition 7. Let $\tilde{\varphi}:[0,1]^{n} \rightarrow \mathbb{R}$ be the multilinear extension of a pseudo-Boolean function $\varphi$ and let
$\mu_{i}\left(\mathbf{x}^{(i)}\right), i=1, \ldots, n$, be pseudo-Boolean functions. Let us consider the pseudo-Boolean function $\mu(\mathbf{x})=\tilde{\varphi}\left(\mu_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \mu_{n}\left(\mathbf{x}^{(n)}\right)\right)$, where $\mathbf{x}=\left(\mathbf{x}^{(1)}, \ldots, \mathbf{x}^{(n)}\right)$. Then the multilinear extension of $\mu$ can be computed as

$$
\tilde{\mu}(\mathbf{x})=\tilde{\varphi}\left(\tilde{\mu}_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \tilde{\mu}_{n}\left(\mathbf{x}^{(n)}\right)\right) .
$$

Remark 3. Proposition 7 allows us to represent the aggregation (5) using more simple aggregations as follows. Let $\tilde{\varphi}\left(t_{1}, \ldots, t_{n}\right)$ be a multilinear extension of a pseudo-Boolean function $\varphi:\{0,1\}^{n} \rightarrow[0,1]$. Then we can consider the following sequence of pseudo-Boolean functions

$$
\begin{gathered}
\varphi_{0}=\tilde{\varphi}\left(t_{1}, \ldots, t_{n}\right), \varphi_{1}=\tilde{\varphi}\left(\mu_{1}\left(\mathbf{x}^{(1)}\right), t_{2}, \ldots, t_{n}\right) \\
\varphi_{2}=\tilde{\varphi}\left(\mu_{1}\left(\mathbf{x}^{(1)}\right), \mu_{2}\left(\mathbf{x}^{(2)}\right), t_{3}, \ldots, t_{n}\right) \\
\varphi_{n}(\mathbf{x})=\tilde{\varphi}\left(\mu_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \mu_{n}\left(\mathbf{x}^{(n)}\right)\right)
\end{gathered}
$$

where $t_{i} \in\{0,1\}, \quad i=1, \ldots, n, \quad$ and corresponding aggregation functions:

$$
\begin{gathered}
\tilde{\varphi}_{0}=\tilde{\varphi}\left(t_{1}, \ldots, t_{n}\right), \tilde{\varphi}_{1}=\tilde{\varphi}\left(\tilde{\mu}_{1}\left(\mathbf{x}^{(1)}\right), t_{2}, \ldots, t_{n}\right) \\
\tilde{\varphi}_{2}=\tilde{\varphi}\left(\tilde{\mu}_{1}\left(\mathbf{x}^{(1)}\right), \tilde{\mu}_{2}\left(\mathbf{x}^{(2)}\right), t_{3}, \ldots, t_{n}\right) \\
\tilde{\varphi}_{n}(\mathbf{x})=\tilde{\varphi}\left(\tilde{\mu}_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \tilde{\mu}_{n}\left(\mathbf{x}^{(n)}\right)\right)
\end{gathered}
$$

that have to be obviously multilinear extensions of corresponding pseudo-Boolean functions. Each $\varphi_{i}$ is generated from $\varphi_{i-1}$ by replacing variable $t_{i}$ with the pseudo-Boolean function $\mu_{i}\left(\mathbf{x}^{(i)}\right)$.
The interpretation of simple aggregations, considered in Remark 3, through set functions is given in the following lemma.
Lemma 2. Let $\varphi_{1}:\{0,1\}^{n} \rightarrow[0,1]$ and $\varphi_{2}:\{0,1\}^{m} \rightarrow[0,1]$ be pseudo-Boolean functions and let $\tilde{\varphi}_{i}, i=1,2$, be their multilinear extensions. Consider their aggregation of the following type:
$\varphi\left(x_{1}, \ldots, x_{n-1}, x_{n+1}, \ldots, x_{n+m}\right)=\tilde{\varphi}_{1}\left(x_{1}, \ldots, x_{n-1}, \varphi_{2}\left(x_{n+1}, \ldots, x_{n+m}\right)\right)$, and corresponding set functions on $2^{Z}$, where $Z=\{1, \ldots, n+m\}$ :

$$
\begin{gathered}
\mu_{1}(A)=\varphi_{1}\left(\mathbf{1}_{A}\right), \text { where } A \subseteq\{1,2, \ldots, n\} ; \\
\mu_{2}(B)=\varphi_{2}\left(\mathbf{1}_{B}\right) \text {, where } B \subseteq\{n+1, \ldots, n+m\} ; \\
\mu(C)=\varphi\left(\mathbf{1}_{C}\right) \text {, where } C \subseteq\{1, \ldots, n-1, n+1, \ldots, n+m\}
\end{gathered}
$$

Then

$$
\mu(A \cup B)=\mu_{1}(A)+\left(\mu_{1}(A \cup\{n\})-\mu_{1}(A)\right) \mu_{2}(B)
$$

where $A \subseteq\{1, \ldots, n-1\}$ and $B \subseteq\{n+1, \ldots, n+m\}$.
Like in the theory of Boolean functions, let us introduce the notion of essential variable for pseudo-Boolean functions. Let $\varphi:\{0,1\}^{n} \rightarrow[0,1]$ be a pseudo-Boolean function. The variable $x_{i}$ is called essential for $\varphi$ if there are vectors $\mathbf{x}_{1}=\left(x_{1}, \ldots, x_{i-1}, 0, x_{i+1}, \ldots, x_{n}\right)$ and $\mathbf{x}_{2}=\left(x_{1}, \ldots, x_{i-1}, 1, x_{i+1}, \ldots, x_{n}\right) \quad$ in $\{0,1\}^{n} \quad$ such that $\varphi\left(\mathbf{x}_{1}\right) \neq \varphi\left(\mathbf{x}_{2}\right)$. It is easy to express such a property using
set functions. Let $\mu(A)=\varphi\left(\mathbf{1}_{A}\right)$, where $A \in 2^{Z}$. Then the variable $x_{i}$ is essential if the set function $v(A)=\mu(A \cup\{i\})-\mu(A)$, where $A \in 2^{Z}$, is not identical to zero.
Proposition 8. Let $\varphi:\{0,1\}^{n} \rightarrow[0,1]$ be a pseudoBoolean function and let $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ be its multilinear extension. Then the variable $x_{i}$ is essential for $\varphi$ iff there is a $\mathbf{x} \in[0,1]^{n}$ such that $\frac{\partial \tilde{\varphi}(\mathbf{x})}{\partial x_{i}} \neq 0$.
Proposition 9. Let $\mu=\tilde{\varphi} \circ \boldsymbol{\mu}$ be the aggregation defined by formula (5), and let $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ be a multilinear extension of a monotone measure $\varphi$. Then representation $\mu=\tilde{\varphi} \circ \boldsymbol{\mu}$ for fixed sets $X_{1}, \ldots, X_{n}$ is defined uniquely iff each variable in $\varphi$ is essential.
In this section we will prove the following result.
Theorem 3. Let $\tilde{\varphi}:[0,1]^{n} \rightarrow[0,1]$ be a multilinear extension of a 2-monotone measure $\varphi$ on $2^{Z}$ and let all variables of $\tilde{\varphi}$ be essential. Let us assume that $\mu_{i}$ are 2monotone measures on $2^{X_{i}}$, where $X_{1}, \ldots, X_{n}$ are mutually disjoint finite nonempty sets. Then $\mu=\tilde{\varphi} \circ \boldsymbol{\mu}$ is an extreme point iff 2-monotone measures $\varphi, \mu_{1}, \ldots, \mu_{n}$ are extreme points too.

## 7 Examples of extreme 2-monotone measures

Let $\mu$ be an extreme 2-monotone measure. Then we call it perfect if it is uniquely defined by a filter $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$, in other words, an extreme measure is not perfect if there is another extreme 2 monotone measure with the same filter $\mathbf{f}$, on which it has positive values. We will describe next the class of such extreme 2-monotone measures.
Let $\mu$ be a set function on $2^{x}$. We say that $\mu$ is additive on a filter $\mathbf{f}$ if
a) $\mu(A)=0$ for any $A \notin \mathbf{f}$;
b) $\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right)$ for any sets $A \cup\left\{x_{i}\right\}, A \cup\left\{x_{j}\right\} \in \mathbf{f}$ such that $x_{i}, x_{j} \notin A$ and $x_{i} \neq x_{j}$.
Lemma 3. Let a set function $\mu$ be additive on a filter $\mathbf{f}$. Consider any $A \in \mathbf{f}$ and $x_{i} \notin A$. Then $\mu\left(A \cup\left\{x_{i}\right\}\right)-$ $\mu(A)=\mu\left(C \cup\left\{x_{i}\right\}\right)-\mu(C)$ for any $C \in \mathbf{f}$ with $C \subseteq A$.
Corollary 2. If the set function $\mu$ is additive on a filter $\mathbf{f}$, then $\mu\left(A \cup\left\{x_{i}\right\}\right)-\mu(A)=\mu\left(C \cup\left\{x_{i}\right\}\right)-\mu(C)$ for any $A, C \in \mathbf{f}$ such that $A, C \subseteq X \backslash\left\{x_{i}\right\}$.
The results formulated in Lemma 3 and Corollary 2 can be better described by the function

$$
v\left(x_{i}\right)=\mu\left(A \cup\left\{x_{i}\right\}\right)-\mu(A)
$$

where $A \in \mathbf{f}$ and $x_{i} \notin A$. Let us notice that $v\left(x_{i}\right)$ does not depend on the choice of $A$. The value $v\left(x_{i}\right)$ is called the weight of $x_{i}$ on filter $\mathbf{f}$ for a set function $\mu$.
Proposition 10. Let a nonnegative set function $\mu$ be additive on a filter $\mathbf{f}$, and $v\left(x_{i}\right) \geq 0$ for all $x_{i} \in X$. Then $\mu$ is 2-monotone.
Proposition 11. Let a nonnegative set function $\mu$ be additive on a filter $\mathbf{f}$. Let us consider the system of sets $2^{X} \backslash \mathbf{f}$ and the set of its maximal elements $\left\{C_{1}, \ldots, C_{k}\right\}$. Then $\mu$ is 2-monotone if $\left\{\bar{C}_{1}, \ldots, \bar{C}_{k}\right\}$ is a covering of $X$.
Let us consider how to construct 2-monotone measures that are additive on a filter. We prove first the following auxiliary lemma.
Lemma 4. Let $\mathbf{f}$ be a filter of the algebra $2^{X}$. Then the system of sets

$$
\mathbf{f}_{0}=\left\{A \mid A \cup\left\{x_{i}\right\}, A \cup\left\{x_{j}\right\} \in \mathbf{f}, x_{i}, x_{j} \notin A\right\}
$$

is also a filter and $\mathbf{f}_{0} \supseteq \mathbf{f}$.
Proposition 12. Let we use the notations from Lemma 4, $A \in \mathbf{f}_{0}, x_{j} \notin A$, and let a set function $\mu$ be additive on the filter $\mathbf{f}$. Then the value $v\left(x_{j}\right)=\mu\left(A \cup\left\{x_{j}\right\}\right)-\mu(A)$ does not depend on the choice of $A \in \mathbf{f}_{0}$.
Corollary 3. Let $A \in \mathbf{f}_{0}, B \supseteq A$, and let $\mu$ be additive on the filter $\mathbf{f}$. Then

$$
\mu(B)=\mu(A)+\sum_{x_{i} \in B \backslash A} v\left(x_{i}\right) .
$$

Corollary 4. Let $A \cap B \in \mathbf{f}_{0}$ for sets $A$ and $B$, and let $\mu$ be additive on the filter $\mathbf{f}$. Then

$$
\mu(A)+\mu(B)=\mu(A \cap B)+\mu(A \cup B)
$$

Proposition 13. Let a set function $\mu$ be additive on the filter $\mathbf{f}$. Then values of $v$ obeys the following system of equations:

$$
\begin{equation*}
\sum_{x_{i} \notin B} v\left(x_{i}\right)=\mu(X) \text { for all } B \in \mathbf{f}_{0} \backslash \mathbf{f}, \tag{6}
\end{equation*}
$$

in addition

$$
\mu(B)=\left\{\begin{array}{cc}
0, & B \notin \mathbf{f},  \tag{7}\\
\mu(X)-\sum_{x_{i} \notin B} v\left(x_{i}\right), & B \in \mathbf{f},
\end{array}\right.
$$

Conversely, each set function $\mu$ obeying equalities (6) and (7) is additive on the filter $\mathbf{f}$.
Remark 4. Solving equations (6) and (7) w.r.t. $v\left(x_{i}\right)$ we can find all set functions that are additive on the filter $\mathbf{f}$, i.e. it is guaranteed that any such function satisfies

1) $\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right)$ for $A \cup\left\{x_{i}\right\}, A \cup\left\{x_{j}\right\} \in \mathbf{f}$ and $A \cap\left\{x_{i}, x_{j}\right\}=\varnothing$;
2) $\mu(A)=0$ for $A \notin \mathbf{f}$.

However, we can not guarantee that $\mu$ is 2 -monotone, because 2 -monotonicity of $\mu$ in this case is equivalent to $v\left(x_{i}\right) \geq 0$ for all $x_{i} \in X$ by Proposition 10 .

Proposition 14. Let $\mu$ be a 2-monotone measure that is additive on the filter $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$. Then $\mu$ is an extreme 2-monotone measure if it is defined uniquely.
Proposition 15. Let the filter $\mathbf{f}$ obey the conditions formulated in Proposition 3. Then if an extreme 2monotone measure $\mu$, which is additive on $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$, exists, then it is perfect.
Proposition 16. Let the filter $\mathbf{f}$ obey the conditions formulated in Proposition 11. Then if an extreme 2monotone measure $\mu$, which is additive on $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}, \quad$ exists, then it is defined uniquely.
Let us consider examples of perfect 2-monotone measures that are additive on filter. A monotone measure is called symmetrical if its values depend only on the cardinality of the corresponding set. The next proposition gives the description of extreme symmetrical 2 monotone measures.
Proposition 17. Let $X=\left\{x_{1}, x_{2}, \ldots, x_{n}\right\}$. Then any symmetrical monotone measure, defined by

$$
\mu_{k}(A)=\left\{\begin{array}{cc}
0, & |A|<k-1 \\
(m-k+1) /(n-k+1), & |A|=m \geq k-1
\end{array}\right.
$$

where $k=2, \ldots, n$, is a perfect extreme 2-monotone measure.
Remark 5. It is easy to show that the set of all symmetrical 2 -monotone measures on $2^{x}$, where $X=\left\{x_{1}, x_{2}, \ldots, x_{n}\right\}$, is convex and the extreme points of it are measures $\mu_{k}, k=1, \ldots, n$. Obviously, $\mu_{1}$ is not an extreme point of $M_{2-\text { mon }}$ if $n=1$, because it is represented as $\mu_{1}=(1 / n) \sum_{k=1}^{n} \eta_{\left\langle\left\langle x_{k}\right\rangle\right\rangle}$.


Figure. 1: A perfect extreme 2-monotone measure that is additive on the filter.

Remark 6. Let $X=\left\{x_{1}, x_{2}, x_{3}\right\}$, then the extreme points of $M_{2-m o n}(X)$ are perfect and they are additive on the filter of their positive values. These measures are $\eta_{\langle A\rangle}$, where $|A|>0$, and the symmetrical measure $\mu_{2}$ for $n=3$. If $X=\left\{x_{1}, x_{2}, x_{3}, x_{4}\right\}$, then extreme points of $M_{2-\text { mon }}(X)$ are not necessarily measures described in Proposition 17. For example, let us consider the filter $\mathbf{f}=\left\langle\left\{x_{1}, x_{2}\right\},\left\{x_{1}, x_{3}\right\},\left\{x_{1}, x_{4}\right\},\left\{x_{2}, x_{3}, x_{4}\right\}\right\rangle$. Let us try to find a 2 -monotone measure $\mu$ that is additive on
$\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$. In this case, $\mathbf{f}_{0}=\left\langle\left\{x_{1}\right\},\left\{x_{2}, x_{3}\right\}\right.$, $\left.\left\{x_{2}, x_{4}\right\},\left\{x_{3}, x_{4}\right\}\right\rangle$, and the function $v\left(x_{i}\right), x_{i} \in X$, should obey the following linear system of equations:

$$
\left\{\begin{array}{c}
v\left(x_{2}\right)+v\left(x_{3}\right)+v\left(x_{4}\right)=1 \\
v\left(x_{1}\right)+v\left(x_{2}\right)=1 \\
v\left(x_{1}\right)+v\left(x_{3}\right)=1 \\
v\left(x_{1}\right)+v\left(x_{4}\right)=1
\end{array}\right.
$$

that has the following unique solution $v\left(x_{1}\right)=2 / 3$, $v\left(x_{2}\right)=v\left(x_{3}\right)=v\left(x_{4}\right)=1 / 3$. After that we can calculate the values of 2 -monotone measure $\mu$ by the formula (7). This measure is depicted on Figure 1. Using Proposition 15 it easy to check that $\mu$ is a perfect extreme 2monotone measure.
Let us consider an example of extreme 2-monotone measure $\mu$, depicted on Figure 2, that is not additive on the filter $\mathbf{f}=\left\{A \in 2^{X} \mid \mu(A)>0\right\}$. Using Corollary 1 , it is easy to show that it is an extreme point of $M_{2-m o n}$. In addition, it is possible to show that $\mu$ is a perfect extreme 2-monotone measure.


Figure. 2: A perfect extreme 2-monotone measure that is not additive on the filter.
It is easy to find extreme 2-monotone measures that are not perfect. Such measures are depicted on Figure 3, with parameters $\alpha, \beta, \gamma$, and $\lambda$, given in Table 1.


Figure 3: Extreme 2-monotone measures that are not perfect.

| No. | $\alpha$ | $\beta$ | $\gamma$ | $\lambda$ |
| :--- | :---: | :---: | :---: | :---: |
| 1. | $2 / 3$ | $1 / 2$ | $1 / 2$ | $1 / 6$ |
| 2. | $2 / 3$ | $1 / 2$ | $1 / 3$ | $1 / 6$ |
| 3. | $1 / 3$ | $1 / 2$ | $1 / 6$ | $1 / 6$ |
| 4. | $1 / 3$ | $1 / 3$ | $1 / 6$ | $1 / 6$ |
| 5. | $5 / 6$ | $1 / 3$ | $2 / 3$ | $1 / 6$ |

Table 1: Values of parameters $\alpha, \beta, \gamma, \lambda$.

## 8 Conclusion

In this paper we give general necessary and sufficient conditions under which 2-monotone measures are
extreme points of $M_{2-m o n}$, describe some important classes of them, and give ways of their generation. As shown by examples, the introduced class of extreme 2 monotone measures, that are additive on filters do not cover all possible extreme 2 -monotone measures, and we cannot generate all possible extreme 2 -monotone measures based on aggregations with the help of multilinear extension. However, this paper can be considered as the first step to the desirable solution. As one can see from the examples, general extreme 2monotone measures have structures that are similar to a structure of extreme 2-monotone measures that are additive on filter and there is a possibility to generalize it. This can be the topic for the future research.

## Appendix

Proof of Proposition 1. We should prove monotonicity, i.e. $\mu\left(A \cup\left\{x_{k}\right\}\right)-\mu(A) \geq 0 \quad$ for $A \in 2^{X}$ and $x_{k} \notin A$. Let us consider a chain of sets $B_{0}=\varnothing, B_{1}=\left\{y_{1}\right\}, B_{1}=\left\{y_{1}, y_{2}\right\}, \ldots$, $B_{m}=\left\{y_{1}, \ldots, y_{m}\right\}$. Then inequalities (2) imply

$$
\begin{gathered}
0 \leq \mu\left(B_{0} \cup\left\{x_{k}\right\}\right)-\mu\left(B_{0}\right) \leq \mu\left(B_{1} \cup\left\{x_{k}\right\}\right)-\mu\left(B_{1}\right) \leq \ldots \leq \\
\mu\left(B_{m} \cup\left\{x_{k}\right\}\right)-\mu\left(B_{m}\right), \text { i.e. } \mu\left(A \cup\left\{x_{k}\right\}\right)-\mu(A) \geq 0 .
\end{gathered}
$$

Proof of Proposition 3. It is necessary to show that (3) is valid for $A \cup\left\{x_{i}\right\} \in \mathbf{f}$ and $A \cup\left\{x_{j}\right\} \notin \mathbf{f}$. Since $\mu\left(A \cup\left\{x_{j}\right\}\right)=0$ and $\mu(A)=0$, this inequality is transformed to

$$
\mu\left(A \cup\left\{x_{i}\right\}\right) \leq \mu\left(A \cup\left\{x_{i}\right\} \cup\left\{x_{j}\right\}\right)
$$

Let us prove that $\mu\left(B \cup\left\{x_{j}\right\}\right)-\mu(B) \geq 0$ for any $B \in \mathbf{f}$ and $x_{j} \notin B$. Since $B \in \mathbf{f}$, then there exists a minimal element $C_{k} \in \mathcal{C}^{c}$ such that $C_{k} \subseteq B$. Let us show first that

$$
\begin{equation*}
\mu\left(C_{k} \cup\left\{x_{j}\right\}\right)-\mu\left(C_{k}\right) \geq 0 \tag{A1}
\end{equation*}
$$

According to the statement of the proposition for $C_{k} \in \mathscr{C}^{*}$ and $x_{j} \notin C_{k}$ there exists $C_{l} \in \mathscr{C}^{\bullet}$ such that $\left\{x_{j}\right\}=C_{l} \backslash C_{k}$. Since $C_{k} \backslash C_{l} \neq \varnothing$, there is some $x_{i} \in C_{k} \backslash C_{l}$, and obviously $C_{l} \subseteq\left(C_{k} \backslash\left\{x_{i}\right\}\right) \cup\left\{x_{j}\right\}$, i.e. $\left(C_{k} \backslash\left\{x_{i}\right\}\right) \cup\left\{x_{j}\right\} \in \mathbf{f}$. Because $\mu$ is 2-monotone on $\mathbf{f}$, we have
$\mu\left(\left(C_{k} \backslash\left\{x_{j}\right\}\right) \cup\left\{x_{i}\right\}\right)+\mu\left(C_{k}\right) \leq \mu\left(C_{k} \backslash\left\{x_{j}\right\}\right)+\mu\left(C_{k} \cup\left\{x_{i}\right\}\right)$. Let us notice that in the last inequality $\mu\left(\left(C_{k} \backslash\left\{x_{j}\right\}\right) \cup\left\{x_{i}\right\}\right)>0$ and $\mu\left(C_{k} \backslash\left\{x_{j}\right\}\right)=0$, therefore, the inequality (A1) is valid.
Let us show next that this inequality is fulfilled for $B$ if $C_{k} \subseteq B$. For this purpose, consider the following chain of sets

$$
B_{0}=C_{k}, B_{1}=C_{k} \cup\left\{x_{i_{1}}\right\}, \ldots, B_{r}=C_{k} \cup\left\{x_{i_{1}}, \ldots, x_{i_{r}}\right\}=B
$$

Since $\mu$ is 2-monotone on the filter $\mathbf{f}$, the following inequalities are valid:

$$
\begin{gathered}
\mu\left(B_{0} \cup\left\{x_{j}\right\}\right)-\mu\left(B_{0}\right) \leq \mu\left(B_{1} \cup\left\{x_{j}\right\}\right)-\mu\left(B_{1}\right) \leq \ldots \\
\leq \mu\left(B_{r} \cup\left\{x_{j}\right\}\right)-\mu\left(B_{r}\right)
\end{gathered}
$$

i.e. $\mu\left(C_{k} \cup\left\{x_{i}\right\}\right)-\mu(C) \leq \mu\left(B \cup\left\{x_{i}\right\}\right)-\mu(B)$.

Proof of Theorem 1. Obviously, $\rho_{\gamma} \supseteq \rho_{\Gamma}$ for any order $\rho_{\gamma}$ with $\gamma \in \Gamma$. Let us show next that if $\rho_{\gamma^{\prime}} \supseteq \rho_{\Gamma}$, then $\gamma^{\prime} \in \Gamma$. For this purpose, it is necessary to show that sets $B_{1}=\left\{y_{1}\right\}$, $B_{2}=\left\{y_{1}, y_{2}\right\}, \quad \ldots, \quad B_{n}=\left\{y_{1}, y_{2}, \ldots, y_{n}\right\}, \quad$ where $\quad y_{i}=x_{\gamma^{\prime}(i)}$, $i=1,2, \ldots, n$, are in $\mathfrak{L}$. Let us show first that $B_{1} \in \mathfrak{L}$. Let us put into correspondence to each permutation $\gamma \in \Gamma$ the set
$B_{\gamma}\left(y_{1}\right)=\left\{x_{\gamma(1)}, x_{\gamma(2)}, \ldots, x_{\gamma(m)}\right\}$ such that $y_{1}=x_{\gamma(m)}$. It is easy to see that conditions $\quad \rho_{\Gamma}=\bigcap_{\gamma \in \Gamma} \rho_{\gamma} \quad$ and $\quad \rho_{\gamma^{\prime}} \supseteq \rho_{\Gamma} \quad$ imply $\bigcap_{\gamma \in \Gamma} B_{\gamma}\left(y_{1}\right)=\left\{y_{1}\right\}$, i.e. $B_{1} \in \mathfrak{L}$. We then prove $B_{k} \in \mathfrak{L}$, $k=2, \ldots, n$, by induction. Let us assume that $B_{1}, \ldots, B_{k-1} \in \mathfrak{L}$ and show that $B_{k} \in \mathfrak{L}$. In this case the conditions $\rho_{\Gamma}=\bigcap_{\gamma \in \Gamma} \rho_{\gamma}$ and $\quad \rho_{\gamma^{\prime}} \supseteq \rho_{\Gamma} \quad$ imply $\bigcap_{\gamma \in \Gamma} B_{\gamma}\left(y_{k}\right) \subseteq B_{k-1} \cup\left\{y_{k}\right\}$. Therefore, $B_{k-1} \cup \bigcap_{\gamma \in \Gamma} B_{\gamma}\left(y_{k}\right)=B_{k}$, i.e. $B_{k} \in \mathfrak{L}$.
Proof of Proposition 5. Necessity. Let us assume that $\mu$ is not an extreme point in $M_{2-m o n}$. Then it can be represented in the form $\mu=a \mu_{1}+(1-a) \mu_{2}$, where $a \in(0,1), \mu_{1}, \mu_{2} \in M_{2-\text { mon }}$ and $\mu_{1} \neq \mu$. Clearly, $\mu_{1}$ obeys conditions on $v 1$ ) and 2) in this proposition.
Sufficiency. Let us assume to the contrary that there exists $v \in M_{2-\text { mon }}$ with properties 1) and 2). We will show that in this case $\mu$ is not an extreme point in $M_{2-m o n}$. For this purpose, let us consider a set function $\theta_{a}(A)=\mu(A)-a \nu(A)$, parametrically depending on $a \in[0,1]$ and also

$$
\begin{gathered}
\varepsilon_{1}=\max \left\{a \in[0,1] \mid \theta_{a}(A) \geq 0 \text { for all } A \in 2^{x}\right\}, \\
\varepsilon_{2}=\max \left\{a \in[0,1] \mid \theta_{a}(A)+\theta_{a}(B) \leq\right. \\
\left.\theta_{a}(A \cap B)+\theta_{a}(A \cup B) \text { for all } A, B \in 2^{x}\right\}
\end{gathered}
$$

It is easy to see that conditions 1) and 2) imply that $\varepsilon_{1}>0$ and $\varepsilon_{2}>0$. Therefore, a set function $\theta_{b}$, where $b=\min \left\{\varepsilon_{1}, \varepsilon_{2}\right\}$, is nonnegative and 2 -monotone. Thus, $\mu$ is represented as

$$
\mu=b v(A)+(1-b) \mu_{2}
$$

where $\mu_{2}=\theta_{b} /(1-b)$ and, obviously, $v, \mu_{2} \in M_{2-\text { mon }}$, i.e. $\mu$ is not an extreme point in $M_{2-\text { mon }}$
Proof of Proposition 7. Clearly, $\tilde{\mu}(\mathbf{x})=\tilde{\varphi}\left(\tilde{\mu}_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \tilde{\mu}_{n}\left(\mathbf{x}^{(n)}\right)\right)$ for every binary vector $\mathbf{x}$ and $\tilde{\varphi}\left(\tilde{\mu}_{1}\left(\mathbf{x}^{(1)}\right), \ldots, \tilde{\mu}_{n}\left(\mathbf{x}^{(n)}\right)\right)$ is a multilinear polynomial. Therefore, the proposition follows from the uniqueness of such a polynomial for the pseudo-Boolean function $\mu$.
Proof of Lemma 2. Using the Taylor decomposition at the point $\mathbf{x}=\left(x_{1}, \ldots, x_{n-1}, 0\right)$, we get

$$
\begin{gathered}
\varphi\left(x_{1}, \ldots, x_{n-1}, x_{n+1}, \ldots, x_{n+m}\right)=\varphi_{1}\left(x_{1}, \ldots, x_{n-1}, 0\right)+ \\
\frac{\partial \tilde{\varphi}_{1}\left(x_{1}, \ldots, x_{n-1}, 0\right)}{\partial x_{n}} \varphi_{2}\left(x_{n+1}, \ldots, x_{n+m}\right)
\end{gathered}
$$

Then we find that if $\left(x_{1}, \ldots, x_{n-1}, x_{n+1}, \ldots, x_{n+m}\right)=\mathbf{1}_{A \cup B}$, then

$$
\varphi_{1}\left(x_{1}, \ldots, x_{n-1}, 0\right)=\mu_{1}(A)
$$

$\frac{\partial \tilde{\varphi}_{1}\left(x_{1}, \ldots, x_{n-1}, 0\right)}{\partial x_{n}}=\mu_{1}(A \cup\{n\})-\mu_{1}(A)$,
$\varphi_{2}\left(x_{n+1}, \ldots, x_{n+m}\right)=\mu_{2}(B)$
Proof of Proposition 8. Let $\mu(A)=\varphi\left(\mathbf{1}_{A}\right)$, where $A \in 2^{Z}$. Then the multilinear extension of $\varphi$ can be represented as

$$
\tilde{\varphi}(\mathbf{x})=\sum_{A \subseteq Z} \mu(A) \prod_{k \in A} x_{k} \prod_{k \notin A}\left(1-x_{k}\right) .
$$

Taking partial derivative, we get

$$
\frac{\partial \tilde{\varphi}(\mathbf{x})}{\partial x_{i}}=\sum_{A \subseteq Z \backslash\{i\}}(\mu(A \cup\{i\})-\mu(A)) \prod_{k \in A} x_{k} \prod_{k \notin A}\left(1-x_{k}\right) .
$$

The proposition follows from the last formula.
Proof of Proposition 9. Let us show that $\varphi$ is defined uniquely. Let $\mathbf{x} \in\{0,1\}^{n}$ and $A=\bigcup_{i=1}^{n} A_{i}$, where $A_{i} \in 2^{X_{i}}$, is chosen such that $A_{i}=X_{i}$ if $x_{i}=1$ and $A_{i}=\varnothing$ if $x_{i}=0$. Then $\left(\mu_{1}\left(A \cap X_{1}\right), \ldots, \mu_{n}\left(A \cap X_{n}\right)\right)=\mathbf{x}$, i.e. $\mu(A)=\varphi(\mathbf{x})$. This means that $\varphi$ is defined uniquely by $\mu$.
Let us show that vector $\boldsymbol{\mu}$ is defined uniquely if each variable $x_{i}$ is essential for $\varphi$. Let us assume that the variable $x_{i}$ is essential for $\varphi$. Then by definition there are vectors $\mathbf{x}_{1}=\left(x_{1}, \ldots, x_{i-1}, 0, x_{i+1}, \ldots, x_{n}\right)$ and $\mathbf{x}_{2}=\left(x_{1}, \ldots, x_{i-1}, 1, x_{i+1}, \ldots, x_{n}\right)$ in $\{0,1\}^{n}$ such that $\varphi\left(\mathbf{x}_{1}\right) \neq \varphi\left(\mathbf{x}_{2}\right)$. Let $A=\bigcup_{k=1}^{n} A_{k}$, where $A_{k} \in 2^{X_{k}}$, such that $A_{k}=X_{k}$ if $x_{k}=1$ and $k \neq i ; A_{k}=\varnothing$ if $x_{k}=0$ and $k \neq i$; and $A_{i}$ is chosen arbitrary in $2^{X_{i}}$. Then using the Taylor decomposition, we get

$$
\begin{gathered}
\mu(A)=\tilde{\varphi}\left(x_{1}, \ldots, x_{i-1}, \mu_{i}\left(A_{i}\right), x_{i+1}, \ldots, x_{n}\right)= \\
\tilde{\varphi}\left(\mathbf{x}_{1}\right)+\mu_{i}\left(A_{i}\right) \frac{\partial \tilde{\varphi}\left(\mathbf{x}_{1}\right)}{\partial x_{i}}
\end{gathered}
$$

Therefore, we can calculate

$$
\mu_{i}\left(A_{i}\right)=\left(\mu(A)-\tilde{\varphi}\left(\mathbf{x}_{1}\right)\right) / \frac{\partial \tilde{\varphi}\left(\mathbf{x}_{1}\right)}{\partial x_{i}}
$$

because $\frac{\partial \tilde{\varphi}\left(\mathbf{x}_{1}\right)}{\partial x_{i}} \neq 0$ according to Proposition 8 . Thus, each set function $\mu_{i}$ is defined uniquely if every variable $x_{i}$ is essential. Let us notice that if $\varphi$ contains a nonessential variable $x_{i}$, then $\tilde{\varphi}$ does not depend on $x_{i}$. This implies that the representation $\mu=\tilde{\varphi} \circ \boldsymbol{\mu}$ is not defined uniquely, since any $\mu_{i}$ has no influence on the result of aggregation.
Proof of Theorem 3. Necessity. Consider 2 possible cases.

1) Let us assume to the contrary that $\varphi$ is not an extreme 2 monotone measure, however, $\mu$ is an extreme 2 -monotone measure. Then $\varphi=a \varphi_{1}+(1-a) \varphi_{2}$, where $a \in(0,1)$ and $\varphi_{1}, \varphi_{2}$ are different 2 -monotone measures on $2^{Z}$. Therefore, $\tilde{\varphi}=a \tilde{\varphi}_{1}+(1-a) \tilde{\varphi}_{2} \quad$ and $\quad \mu=\tilde{\varphi} \circ \boldsymbol{\mu}=a \tilde{\varphi}_{1} \circ \boldsymbol{\mu}+(1-a) \tilde{\varphi}_{2} \circ \boldsymbol{\mu}$, where $\tilde{\varphi}_{1} \circ \boldsymbol{\mu}, \tilde{\varphi}_{2} \circ \boldsymbol{\mu}$ are different 2-monotone measures by Proposition 9. But this contradicts our assumption that $\mu$ is an extreme 2-monotone measure.
2) Let us assume to the contrary that $\mu_{i}$ is not an extreme 2monotone measure for some $i \in\{1, \ldots, n\}$, however, $\mu$ is an extreme 2-monotone measure. Then $\mu_{i}$ can be represented as a convex sum of two different 2-monotone measures: $\mu_{i}=a \mu_{i}^{(1)}+(1-a) \mu_{i}^{(2)}$, where $a \in(0,1)$, therefore,

$$
\begin{gathered}
\mu=\tilde{\varphi} \circ\left(\mu_{1}, \ldots, \mu_{i-1}, a \mu_{i}^{(1)}+(1-a) \mu_{i}^{(2)}, \mu_{i+1} \ldots, \mu_{n}\right)= \\
a \tilde{\varphi} \circ\left(\mu_{1}, \ldots, \mu_{i-1}, \mu_{i}^{(1)}, \mu_{i+1} \ldots, \mu_{n}\right)+ \\
(1-a) \tilde{\varphi} \circ\left(\mu_{1}, \ldots, \mu_{i-1}, \mu_{i}^{(2)}, \mu_{i+1} \ldots, \mu_{n}\right),
\end{gathered}
$$

where
$\tilde{\varphi} \circ\left(\mu_{1}, \ldots, \mu_{i-1}, \mu_{i}^{(1)}, \mu_{i+1} \ldots, \mu_{n}\right), \tilde{\varphi} \circ\left(\mu_{1}, \ldots, \mu_{i-1}, \mu_{i}^{(2)}, \mu_{i+1} \ldots, \mu_{n}\right)$,
are different 2 -monotone measures. But this contradicts the assumption that $\mu$ is an extreme 2 -monotone measure.
Sufficiency. We will prove sufficiency by induction. According to Remark 3 any aggregation (5) can be represented as a composition of simple aggregations described in Lemma 2. Therefore, if we prove that any simple aggregation of a type $\varphi\left(x_{1}, \ldots, x_{n-1}, x_{n+1}, \ldots, x_{n+m}\right)=\tilde{\varphi}_{1}\left(x_{1}, \ldots, x_{n-1}, \varphi_{2}\left(x_{n+1}, \ldots, x_{n+m}\right)\right)$, whe re the corresponding set functions $\mu_{1}$ and $\mu_{2}$ are extreme 2monotone measures (see notations from Lemma 2), generates the extreme 2 -monotone measure $\mu$. Then we can also say that the general aggregation produces the extreme 2 -monotone measure if the conditions of the theorem are fulfilled. Let us assume to the contrary that $\mu$ is not an extreme 2 -monotone measure. Then there are 2 different 2 -monotone measures $\mu^{(0)}$ and $\mu^{(1)}$ such that

$$
\mu=a \mu^{(0)}+(1-a) \mu^{(1)}, \text { where } a \in(0,1)
$$

Let us consider 2-monotone measures, generated by a mapping

$$
\psi(i)=\left\{\begin{array}{lc}
i, & i \in\{1, \ldots, n-1\} \\
n, & i \in\{n+1, \ldots, m\}
\end{array}\right.
$$

Obviously, $\mu^{\psi /}=a\left(\mu^{(0)}\right)^{\psi}+(1-a)\left(\mu^{(1)}\right)^{\psi / 2}, \mu^{\psi /}=\mu_{1}$, and $\mu^{(0)}$, $\mu^{(1)}$ are 2 -monotone measures. But according to our assumption $\mu_{1}$ is an extreme 2-monotone measure. Therefore, this implies that $\left(\mu^{(0)}\right)^{\psi}=\mu_{1}$.
Our next step is to show that if $\mu_{2}$ is also an extreme 2 monotone measure, then $\mu^{(0)}=\mu^{(1)}=\mu$.
By Lemma 2, $\mu$ can be represented as
$\mu(A \cup B)=\mu_{1}(A)+\left(\mu_{1}(A \cup\{n\})-\mu_{1}(A)\right) \mu_{2}(B)$,
where $A \subseteq\{1, \ldots, n-1\}$ and $B \subseteq\{n+1, \ldots, n+m\}$. Let us denote $Y=\{n+1, \ldots, n+m\}$. Then, taking in account the correspondence between pseudo-Boolean and set functions, the formula (A2) can be rewritten as

$$
\mu(A \cup B)=\mu(A)+(\mu(A \cup Y)-\mu(A)) \mu_{2}(B)
$$

and we can calculate

$$
\mu_{2}(B)=\frac{\mu(A \cup B)-\mu(A)}{\mu(A \cup Y)-\mu(A)}
$$

for any $A \subseteq\{1, \ldots, n-1\}$ such that $\mu(A \cup Y)-\mu(A)>0$. Let us consider set functions:

$$
\mu_{2}^{(i)}(B)=\frac{\mu^{(i)}(A \cup B)-\mu^{(i)}(A)}{\mu^{(i)}(A \cup Y)-\mu^{(i)}(A)}, i=1,2
$$

of $B \subseteq Y$ for any $A \subseteq\{1, \ldots, n-1\}$ with $\mu(A \cup Y)-\mu(A)>0$ and $\mu^{(i)}(A \cup Y)=\mu(A \cup Y)$. It is easy to show that these set functions are 2 -monotone. Let us notice that we have proved that $\mu^{(i)}(A \cup Y)=\mu(A \cup Y)$ and $\mu^{(i)}(A)=\mu(A)$. After that we easily derive that

$$
a \mu_{2}^{(0)}(B)+(1-a) \mu_{2}^{(1)}(B)=\mu_{2}(B)
$$

By our assumption, $\mu_{2}$ is an extreme 2-monotone measure. This implies that $\mu_{2}^{(0)}=\mu_{2}^{(1)}=\mu_{2}$. Thus, we can write

$$
\mu^{(i)}(A \cup B)=\mu(A)+(\mu(A \cup Y)-\mu(A)) \mu_{2}(B)=\mu(A \cup B)
$$

[^1]for any $A \subseteq\{1, \ldots, n-1\}$ and $B \subseteq\{n+1, \ldots, n+m\}$, i.e. $\mu^{(0)}=\mu^{(1)}=\mu$, but this contradicts our assumption that measures $\mu_{2}^{(0)}$ and $\mu_{2}^{(1)}$ are different.
Proof of Lemma 3. Let us consider the sequence of sets
$$
B_{0}=C, B_{1}=C \cup\left\{x_{i_{1}}\right\}, \ldots, B_{m}=C \cup\left\{x_{i_{1}}, \ldots, x_{i_{m}}\right\}=A .
$$

Since $\mu$ is additive on the filter $\mathbf{f}$, we can write

$$
\begin{gathered}
\mu\left(C \cup\left\{x_{j}\right\}\right)-\mu(C)=\mu\left(B_{1} \cup\left\{x_{i}\right\}\right)-\mu\left(B_{1}\right)=\ldots \\
=\mu\left(B_{m} \cup\left\{x_{i}\right\}\right)-\mu\left(B_{m}\right)
\end{gathered}
$$

i.e. $\mu\left(C \cup\left\{x_{i}\right\}\right)-\mu(C)=\mu\left(A \cup\left\{x_{i}\right\}\right)-\mu(A)$. Thus, the
required equality is valid.
Proof of Corollary 2. By Lemma 3

$$
\mu(X)-\mu\left(X \backslash\left\{x_{i}\right\}\right)=\mu\left(C \cup\left\{x_{i}\right\}\right)-\mu(C)
$$

for any $C \in \mathbf{f}$ with $C \subseteq X \backslash\left\{x_{i}\right\}$. This implies the result.
Proof of Proposition 10. Let us check inequality (3), considering the following possible cases:
a) if $A \cup\left\{x_{i}\right\} \in \mathbf{f}$ and $A \cup\left\{x_{j}\right\} \in \mathbf{f}$, then the inequality (3) follows from the additivity of $\mu$ on $\mathbf{f}$;
b) if $A \cup\left\{x_{i}\right\} \in \mathbf{f}$ and $A \cup\left\{x_{j}\right\} \notin \mathbf{f}$, then (3) is transformed to

$$
\mu\left(A \cup\left\{x_{i}\right\}\right) \leq \mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right)
$$

The last inequality is valid, because according to our assumption $v\left(x_{j}\right) \geq 0$;
c) if $A \cup\left\{x_{i}\right\} \notin \mathbf{f}$ and $A \cup\left\{x_{j}\right\} \notin \mathbf{f}$, then inequality (3) is obviously true.
Proof of Proposition 11. It is sufficient to show that

$$
\begin{equation*}
\mu\left(A \cup\left\{x_{i}\right\}\right)-\mu(A) \geq 0 \tag{A3}
\end{equation*}
$$

for all $A \in \mathbf{f}$ and any $x_{i} \in X$. By the assumption $\left\{\bar{C}_{1}, \ldots, \bar{C}_{k}\right\}$ is a covering of $X$, therefore, there is a set $\bar{C}_{l}$ such that $x_{i} \in \bar{C}_{l}$. Let us consider 2 possible cases.
If $\left|\bar{C}_{l}\right|=1$, i.e. $\bar{C}_{l}=\left\{x_{i}\right\}$, then $x_{i} \in A$ for all $A \in \mathbf{f}$. Obviously, in this case the inequality (A3) is valid.
If $\left|\bar{C}_{l}\right| \geq 2$, then there is $x_{j} \in \bar{C}_{l}$ such that $x_{j} \neq x_{i}$. Since $C_{l}$ is a maximal element in $2^{X} \backslash \mathbf{f}$, then $x_{i} \cup C_{l} \in \mathbf{f}, x_{j} \cup C_{l} \in \mathbf{f}$, and additivity of $\mu$ on $\mathbf{f}$ implies

$$
\begin{aligned}
& v\left(x_{i}\right)=\mu\left(C_{l} \cup\left\{x_{i}, x_{j}\right\}\right)-\mu\left(C_{l} \cup\left\{x_{j}\right\}\right)= \\
& \mu\left(C_{l} \cup\left\{x_{i}\right\}\right)-\mu\left(C_{l}\right)=\mu\left(C_{l} \cup\left\{x_{i}\right\}\right) \geq 0
\end{aligned}
$$

i.e. the inequality (A3) is valid for all $A \in \mathbf{f}$.

Proof of Lemma 4. Clearly $\mathbf{f}_{0} \supseteq \mathbf{f}$. Let us show that $B \in \mathbf{f}_{0}$ and $B \subseteq C$ implies $C \in \mathbf{f}_{0}$. It is sufficient to consider the case, when $B \notin \mathbf{f}$ and $C \notin \mathbf{f}$. Then there exist $x_{i}, x_{j} \notin B$ such that

$$
B=\left(B \cup\left\{x_{i}\right\}\right) \cap\left(B \cup\left\{x_{j}\right\}\right)
$$

By our assumption $C \notin \mathbf{f}$, therefore $x_{i}, x_{j} \notin C$. This implies that $C=\left(C \cup\left\{x_{i}\right\}\right) \cap\left(C \cup\left\{x_{j}\right\}\right)$, i.e. $C \in \mathbf{f}_{0}$.
Proof of Proposition 12. It is necessary to show that $\mu\left(A \cup\left\{x_{i}\right\}\right)-\mu(A)=\nu\left(x_{i}\right)$ for any $A \in \mathbf{f}_{0}$ and $x_{j} \notin A$. Let us show first that if $A \in \mathbf{f}_{0}$, then

$$
\mu(A)+\sum_{x_{i} \notin A} v\left(x_{i}\right)=\mu(X) .
$$

Let us consider two possible cases. Let $A \in \mathbf{f}$ and
$X \backslash A=\left\{y_{1}, y_{2}, \ldots, y_{m}\right\}$. Then

$$
\begin{gathered}
\mu\left(A \cup\left\{y_{1}\right\}\right)=\mu(A)+v\left(y_{1}\right), \\
\mu\left(A \cup\left\{y_{1}, y_{2}\right\}\right)=\mu\left(A \cup\left\{y_{1}\right\}\right)+v\left(y_{2}\right) \\
=\mu(A)+v\left(y_{1}\right)+v\left(y_{2}\right), \\
\vdots \\
\mu(X)=\mu(A)+\sum_{i=1}^{m} v\left(y_{i}\right)
\end{gathered}
$$

i.e. the required equality is valid for $A \in \mathbf{f}$. Let us consider the case, when $A \in \mathbf{f}_{0} \backslash \mathbf{f}$. Then there exist $x_{i}, x_{j} \notin A$ such that $A \cup\left\{x_{i}\right\}, A \cup\left\{x_{j}\right\} \in \mathbf{f}$, and

$$
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right)+\mu(A),
$$

i.e.

$$
\begin{gathered}
\mu\left(A \cup\left\{x_{j}\right\}\right)-\mu(A)=\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right)- \\
\mu\left(A \cup\left\{x_{i}\right\}\right)=v\left(x_{j}\right) .
\end{gathered}
$$

After that we see that $\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(A)+v\left(x_{j}\right)$ and

$$
\mu(X)=\mu\left(A \cup\left\{x_{j}\right\}\right)+\sum_{x_{i} \notin A \cup\left\{x_{j}\right\}} v\left(x_{i}\right)=\mu(A)+\sum_{x_{i} \notin A} v\left(x_{i}\right) .
$$

Thus, we can write

$$
\begin{gathered}
\mu\left(A \cup\left\{x_{j}\right\}\right)-\mu(A)=\mu(X)-\sum_{x_{i} \notin A \cup\left\{x_{j}\right\}} v\left(x_{i}\right)- \\
\left(\mu(X)-\sum_{x_{i} \notin A} v\left(x_{i}\right)\right)=v\left(x_{j}\right) .
\end{gathered}
$$

## Proof of Corollary 4.

$$
\begin{gathered}
\mu(A)+\mu(B)=\mu(A \cap B)+\sum_{x_{i} \in A \backslash B} v\left(x_{i}\right)+\mu(A \cap B)+ \\
\sum_{x_{i} \in B \backslash A} v\left(x_{i}\right)=\mu(A \cap B)+\mu(A \cap B)+ \\
\sum_{x_{i} \in(A \cup B) \backslash(A \cap B)} v\left(x_{i}\right)=\mu(A \cap B)+\mu(A \cup B) .
\end{gathered}
$$

Proof of Proposition 13. The first part of the proposition follows from the results considered above. Let us prove the second part. For this purpose, let us show that any set function $\mu$, obeying (6) and (7) is additive on $\mathbf{f}$, i.e.

$$
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right),
$$

for $A \cup\left\{x_{i}\right\}, A \cup\left\{x_{j}\right\} \in \mathbf{f}$ and $A \cap\left\{x_{i}, x_{j}\right\}=\varnothing$.
Let us consider 2 possible cases. Let $A \in \mathbf{f}$, then

$$
\begin{gathered}
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(X)- \\
\sum_{x_{k} \notin A \cup\left\{x_{i}\right\}} v\left(x_{k}\right)+\mu(X)-\sum_{x_{k} \notin A \cup\left\{x_{i}\right\}} v\left(x_{k}\right)= \\
\mu(X)-\sum_{x_{k} \notin A} v\left(x_{k}\right)+\mu(X)-\sum_{x_{k} \notin A \cup\left\{x_{i}, x_{j}\right\}} v\left(x_{k}\right)= \\
\mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right) .
\end{gathered}
$$

Let $A \in \mathbf{f}_{0} \backslash \mathbf{f}$, then

$$
\begin{gathered}
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(X)- \\
\sum_{x_{k} \notin A \cup\left\{x_{i}\right\}} v\left(x_{k}\right)+\mu(X)-\sum_{x_{k} \notin A \cup\left\{x_{i}\right\}} v\left(x_{k}\right)= \\
\mu(X)-\sum_{x_{k} \notin A} v\left(x_{k}\right)+\mu(X)-\sum_{x_{k} \notin A \cup\left\{x_{i}, x_{j}\right\}} v\left(x_{k}\right) .
\end{gathered}
$$

In the last expression $\sum_{x_{k} \notin A} v\left(x_{k}\right)=\mu(X)$, in addition, $\mu(A)=0$. This implies that

$$
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right) .
$$

Proof of Proposition 14. Let us assume to the contrary that $\mu$ is uniquely defined by the filter, but it is not extreme. Then there are 2 different 2-monotone measures $\mu_{1}$ and $\mu_{2}$ such that
$\mu=a \mu_{1}+(1-a) \mu_{2}$, where $a \in(0,1)$. It easy to check that both measures $\mu_{1}$ and $\mu_{2}$ are additive on filter $\mathbf{f}$ but this contradicts our assumption.
Proof Proposition 15. Let us assume to the contrary that $\mu$ obeys conditions of the proposition, however, there is another extreme 2-monotone measure $v$ with $\mathbf{f}=\left\{A \in 2^{X} \mid v(A)>0\right\}$. Let us consider the set function $\theta_{a}(A)=v(A)-a \mu(A)$, parametrically depending on $a \in[0,1]$ and

$$
b=\max \left\{a \in[0,1] \mid \theta_{a}(A) \geq 0 \text { for all } A \in 2^{x}\right\}
$$

Clearly, $b>0$ and the set function $\theta_{b}$ is 2-monotone on the filter $\mathbf{f}$. According to Proposition $3 \theta_{b}$ is 2-monotone on $2^{X}$. Therefore, we can represent $v$ as

$$
v=b \mu+(1-b) \mu_{2}
$$

where $\mu_{2}=\theta_{b} /(1-b)$ is a 2 -monotone measure, but this contradicts our assumption.
Proof of Proposition 16. Let us assume to the contrary that $\mu$ obeys conditions of the proposition, however, there is another extreme 2 -monotone measure $v$, which is additive on $\mathbf{f}=\left\{A \in 2^{X} \mid v(A)>0\right\}$. Let us consider the set function $\theta_{a}(A)=\mu(A)-a v(A)$, parametrically depending on $a \in[0,1]$ and

$$
b=\max \left\{a \in[0,1] \mid \theta_{a}(A) \geq 0 \text { for all } A \in 2^{x}\right\}
$$

Clearly, $b>0$ and the set function $\theta_{b}$ is additive on the filter f. According to Proposition 11, $\theta_{b}$ is 2-monotone on $2^{X}$. Therefore, we can represent $\mu$ as

$$
\mu=b v+(1-b) \mu_{2}
$$

where $\mu_{2}=\theta_{b} /(1-b)$ is a 2-monotone measure, but this contradicts our assumption.
Proof of Proposition 17. Let us notice that $\mu_{n}=\eta_{\langle X\rangle}$ and for this case the proposition is obviously true. Let us check that $\mu_{k}$, where $k \in\{2, \ldots, n-1\}$, is additive on the filter $\mathbf{f}=\left\{A \in 2^{X}| | A \mid \geq k\right\}$, i.e. the following equality holds

$$
\mu\left(A \cup\left\{x_{i}\right\}\right)+\mu\left(A \cup\left\{x_{j}\right\}\right)=\mu(A)+\mu\left(A \cup\left\{x_{i}, x_{j}\right\}\right),(\mathrm{A} 4)
$$ for $A \cup\left\{x_{i}\right\}, A \cup\left\{x_{j}\right\} \in \mathbf{f} \quad$ and $A \cap\left\{x_{i}, x_{j}\right\}=\varnothing$. Let $|A|=m \geq k-1$, then the equality (A4) is transformed to

$$
\frac{m-k+2}{n-k+1}+\frac{m-k+2}{n-k+1}=\frac{m-k+1}{n-k+1}+\frac{m-k+3}{n-k+1}
$$

i.e. (A4) is valid for this case.

Let us show that $\mu_{k}$ is an extreme 2 -monotone measure. In this case $\mathbf{f}_{0}=\left\{A \in 2^{X}| | A \mid \geq k-1\right\}$ and by Proposition 13 all possible 2-monotone measures that are additive on $\mathbf{f}$ can be found by solving the following linear system of equations:

$$
\sum_{x_{i} \notin A} v\left(x_{i}\right)=1 \text { for all } A \in 2^{X} \text { with }|A|=k-1 .
$$

It is easy to check that the solution is uniquely defined by $v\left(x_{i}\right)=1 /(n-k+1), \quad i=1, \ldots, n$. This implies that $\mu_{k}$ is an extreme 2 -monotone measure. It is easy to check that $\mu_{k}$ is a perfect extreme 2-monotone measure by Proposition 15.

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[^0]:    1 Here is used the usual intersection of relations, i.e. if $\rho_{1}, \rho_{2} \in\{1, \ldots, n\} \times\{1, \ldots, n\}$, then $\rho_{1} \cap \rho_{2}$ is the usual intersection defined for sets.

[^1]:    ${ }^{2} \mu^{\psi /}$ denotes a measure on $2^{\{1, \ldots, n\}}$ such that $\mu^{\psi /}(A)=$ $\mu\left(\psi^{-1}(A)\right)$, where $\psi^{-1}(A)=\{i \in\{1, \ldots, n+m\} \mid \psi(i) \in A\}$.

