# Reliability analysis and lattice polynomial system representation 

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## System

Definition. A system consists of several interconnected units

## Assumptions:

(1) The system and the units are of the crisply on/off kind
(2) A serially connected segment of units is functioning if and only if every single unit is functioning

(3) A system of parallel units is functioning if and only at least one unit is functioning


## System

Example. Home video system

1. Blu-ray player
2. DVD player
3. LCD monitor
4. Amplifier
5. Speaker A
6. Speaker B


## Structure function

## Definition.

The state of a component $i \in[n]=\{1, \ldots, n\}$ can be represented by a Boolean variable

$$
x_{i}= \begin{cases}1 & \text { if component } i \text { is functioning } \\ 0 & \text { if component } i \text { is in a failed state }\end{cases}
$$

The state of the system is described from the component states through a Boolean function $\phi:\{0,1\}^{n} \rightarrow\{0,1\}$

$$
\phi\left(x_{1}, \ldots, x_{n}\right)= \begin{cases}1 & \text { if the system is functioning } \\ 0 & \text { if the system is in a failed state }\end{cases}
$$

This function is called the structure function of the system

## Structure function

Series structure


Parallel structure


$$
\phi(\mathbf{x})=1-\left(1-x_{1}\right)\left(1-x_{2}\right)\left(1-x_{3}\right)=\coprod_{i=1}^{3} x_{i}
$$

## Structure function

Home video system


## Coherent and semicoherent systems

## Definition.

Let $\phi:\{0,1\}^{n} \rightarrow\{0,1\}$ be a structure function on $[n]=\{1, \ldots, n\}$.
The system is said to be semicoherent if

- $\phi$ is nondecreasing : $\mathbf{x} \leqslant \mathbf{x}^{\prime} \Rightarrow \phi(\mathbf{x}) \leqslant \phi\left(\mathbf{x}^{\prime}\right)$
- $\phi(\mathbf{0})=0, \phi(\mathbf{1})=1$

The system is said to be coherent if, in addition

- every component is relevant to $\phi$ :

$$
\exists \mathbf{x} \in\{0,1\}^{n}: \phi\left(1_{i}, \mathbf{x}\right) \neq \phi\left(0_{i}, \mathbf{x}\right)
$$

where

$$
\begin{array}{r}
\left(1_{i}, \mathbf{x}\right)=\left(x_{1}, \ldots, \stackrel{(i)}{1}, \ldots, x_{n}\right) \\
\left(0_{i}, \mathbf{x}\right)=\left(x_{1}, \ldots, 0, \ldots, x_{n}\right)
\end{array}
$$

## Representations of Boolean functions

$$
\left.\begin{array}{rl}
\begin{array}{l}
\text { Boolean function } \\
\phi:\{0,1\}^{n} \rightarrow\{0,1\}
\end{array} & \longleftrightarrow
\end{array} \begin{array}{l}
\text { set function } \\
\\
v: 2^{[n]} \rightarrow\{0,1\}
\end{array}\right] \begin{array}{cl}
v(A)=\phi\left(\mathbf{e}_{A}\right) \quad A \subseteq[n]
\end{array}
$$

$\rightarrow$ We write $\phi_{v}$ instead of $\phi$

Representations of a Boolean function

$$
\phi_{v}(\mathbf{x})=\sum_{A \subseteq[n]} v(A) \prod_{i \in A} x_{i} \prod_{i \in[n] \backslash A}\left(1-x_{i}\right)
$$

## Representations of Boolean functions

## Alternative representations

$$
\phi_{v}(\mathbf{x})=\sum_{A \subseteq[n]} m_{v}(A) \prod_{i \in A} x_{i}
$$

where

$$
m_{v}(A)=\sum_{B \subseteq A}(-1)^{|A|-|B|} v(B)
$$

If $\phi_{v}$ is nondecreasing and nonconstant:

$$
\phi_{v}(\mathbf{x})=\coprod_{\substack{A \subseteq[n] \\ v(A)=1}} \prod_{i \in A} x_{i}
$$

(Hammer and Rudeanu 1968)

## System and component lifetimes

Any component $i \in[n]$ has a random lifetime : $T_{i}$ The system has a random lifetime: $T_{S}$

The structure function induces a functional relationship between the variables $T_{1}, \ldots, T_{n}$ and the variable $T_{S}$

## Example:



$$
\begin{gathered}
\phi(\mathbf{x})=x_{1} x_{2} x_{3}=\prod_{i=1}^{3} x_{i} \\
T_{S}=T_{1} \wedge T_{2} \wedge T_{3}=\bigwedge_{i=1}^{3} T_{i}
\end{gathered}
$$

## System and component lifetimes

Home video system


In general,

$$
T_{S}=p\left(T_{1}, \ldots, T_{n}\right)
$$

where $p: \mathbb{R}_{+}^{n} \rightarrow \mathbb{R}_{+}$is an $n$-ary lattice polynomial function
$\Rightarrow$ Formal parallelism between two representations of systems: structure functions and lattice polynomial functions

## Lattice polynomial functions

Let $L \subseteq[-\infty, \infty]$ a totally ordered bounded lattice
$\Rightarrow \wedge=\min$ and $\vee=\max$

The class of n-ary lattice polynomial (I.p.) functions is defined as follows:
(i) For any $k \in[n]$, the projection $\left(t_{1}, \ldots, t_{n}\right) \mapsto t_{k}$ is an $n$-ary I.p. function
(ii) If $p$ and $q$ are $n$-ary l.p. functions then $p \wedge q$ and $p \vee q$ are $n$-ary l.p. functions
(iii) Every $n$-ary I.p. function is constructed by finitely many applications of the rules (i) and (ii).

Example:

$$
p\left(t_{1}, t_{2}, t_{3}\right)=\left(t_{1} \wedge t_{2}\right) \vee t_{3}
$$

## Lattice polynomial functions

Let $a=\inf L$ and $b=\sup L$

$$
\begin{array}{ll}
\text { I.p. function } \longleftrightarrow & \text { set function } \\
p: L^{n} \rightarrow L & w: 2^{[n]} \rightarrow\{a, b\}
\end{array}
$$

$$
w(A)=p\left(\mathbf{e}_{A}^{a, b}\right) \quad A \subseteq[n]
$$

Example : $\mathbf{e}_{\{1,2\}}^{a, b}=(b, b, a, \ldots, a)$ $\rightarrow$ We write $p_{w}$ instead of $p$

Representations of an I.p. function (Birkhoff 1967)

$$
p_{w}(\mathbf{t})=\bigvee_{\substack{A \subseteq[n] \\ w(A)=b}} \bigwedge_{i \in A} t_{i}
$$

## Formal parallelism between the two representations

$T_{i}=$ random lifetime of component $i \in[n]$ $X_{i}(t)=\operatorname{Ind}\left(T_{i}>t\right)=$ random state of $i$ at time $t \geqslant 0$


$$
X_{i}(t)= \begin{cases}1 & \text { if } i \text { is functioning at time } t \\ 0 & \text { if } i \text { is in a failed state at time } t\end{cases}
$$

For the system:
$T_{S}=$ system lifetime
$X_{S}(t)=\operatorname{Ind}\left(T_{S}>t\right)=$ random state of the system at time $t \geqslant 0$

## Formal parallelism between the two representations

Home video system

$$
\begin{gathered}
p_{w}(\mathbf{T})=\left(T_{1} \vee T_{2}\right) \wedge T_{3} \wedge T_{4} \wedge\left(T_{5} \vee T_{6}\right) \\
\phi_{v}(\mathbf{X}(t))=\left(X_{1}(t) \amalg X_{2}(t)\right) X_{3}(t) X_{4}(t)\left(X_{5}(t) \amalg X_{6}(t)\right)
\end{gathered}
$$

$\phi_{v}$ is also an l.p. function that has just the same max-min form as $p_{w}$ but applied to binary arguments

$$
\begin{gathered}
\phi_{v} \longleftrightarrow p_{w} \\
w=\gamma \circ v \\
\gamma:\{0,1\} \rightarrow\{a, b\}, \gamma(0)=a, \gamma(1)=b
\end{gathered}
$$

As the lifetimes are $[0, \infty]$-valued, we now assume that $a=0$ and $b=\infty$

## Formal parallelism between the two representations

## Theorem. (Dukhovny and M. 2008)

Consider a system whose structure function $\phi_{v}:\{0,1\}^{n} \rightarrow\{0,1\}$ is nondecreasing and nonconstant. Then we have

$$
\begin{equation*}
T_{S}=p_{w}\left(T_{1}, \ldots, T_{n}\right) \tag{1}
\end{equation*}
$$

where $w=\gamma \circ v$. Conversely, any system fulfilling (1) for some l.p. function $p_{w}: L^{n} \rightarrow L$ has the nondecreasing and nonconstant structure function $\phi_{v}$, where $v=\gamma^{-1} \circ w$

The proof mainly lies on the immediate identities

$$
\begin{aligned}
\operatorname{Ind}\left(E \wedge E^{\prime}\right) & =\operatorname{Ind}(E) \wedge \operatorname{Ind}\left(E^{\prime}\right) \\
\operatorname{Ind}\left(E \vee E^{\prime}\right) & =\operatorname{Ind}(E) \vee \operatorname{Ind}\left(E^{\prime}\right)
\end{aligned}
$$

valid for all events $E$ and $E^{\prime}$

## Formal parallelism between the two representations

Proof. For every $t \geqslant 0$ we have

$$
\begin{aligned}
\phi_{v}(\mathbf{X}(t)) & =\coprod_{\substack{A \subseteq[n] \\
v(A)=1}} \prod_{i \in A} X_{i}(t) \\
& =\bigvee_{\substack{A \subseteq[n] \\
v(A)=1}} \bigwedge_{i \in A} \operatorname{Ind}\left(T_{i}>t\right)=\operatorname{Ind}\left(\bigvee_{\substack{A \subseteq[n] \\
v(A)=1}} \bigwedge_{i \in A} T_{i}>t\right) \\
& =\operatorname{Ind}\left(p_{w}(\mathbf{T})>t\right)
\end{aligned}
$$

Hence, we have

$$
\begin{array}{ll} 
& X_{S}(t)=\phi_{V}(\mathbf{X}(t)) \quad \forall t \geqslant 0 \\
\Leftrightarrow & \operatorname{Ind}\left(T_{S}>t\right)=\operatorname{Ind}\left(p_{w}(\mathbf{T})>t\right) \quad \forall t \geqslant 0 \\
\Leftrightarrow & T_{S}=p_{w}(\mathbf{T})
\end{array}
$$

## Advantages of the lattice polynomial language

Properties of I.p. functions reveal properties of structure functions

Example. Any I.p. function $p: L^{n} \rightarrow L$ satisfies trivially the functional equations

$$
\begin{array}{ll}
p\left(u \wedge t_{1}, \ldots, u \wedge t_{n}\right)=u \wedge p\left(t_{1}, \ldots, t_{n}\right) & \forall u \in L \\
p\left(u \vee t_{1}, \ldots, u \vee t_{n}\right)=u \vee p\left(t_{1}, \ldots, t_{n}\right) & \forall u \in L
\end{array}
$$

The corresponding equations for the structure functions are

$$
\begin{aligned}
\phi\left(y x_{1}, \ldots, y x_{n}\right) & =y \phi\left(x_{1}, \ldots, x_{n}\right) & & \forall y \in\{0,1\} \\
\phi\left(y \amalg x_{1}, \ldots, y \amalg x_{n}\right) & =y \amalg \phi\left(x_{1}, \ldots, x_{n}\right) & & \forall y \in\{0,1\}
\end{aligned}
$$

## Advantages of the lattice polynomial language

Properties of structure functions reveal properties of I.p. functions

Example. Pivotal decomposition of the structure function

$$
\phi(\mathbf{x})=x_{i} \phi\left(1_{i}, \mathbf{x}\right)+\left(1-x_{i}\right) \phi\left(0_{i}, \mathbf{x}\right)
$$

Bridge structure


$$
\begin{gathered}
\phi(\mathbf{x})=x_{3} \phi\left(1_{3}, \mathbf{x}\right)+\left(1-x_{3}\right) \phi\left(0_{3}, \mathbf{x}\right) \\
\phi\left(1_{3}, \mathbf{x}\right)=\left(x_{1} \amalg x_{2}\right)\left(x_{4} \amalg x_{5}\right) \\
\phi\left(0_{3}, \mathbf{x}\right)=\left(\begin{array}{ll}
x_{1} & \left.x_{4}\right) \amalg\left(x_{2} x_{5}\right)
\end{array}\right.
\end{gathered}
$$

## Advantages of the lattice polynomial language

Corresponding property of the l.p. functions ?

$$
p(\mathbf{t})=\operatorname{median}\left(p\left(a_{i}, \mathbf{t}\right), t_{i}, p\left(b_{i}, \mathbf{t}\right)\right)
$$

where

$$
\operatorname{median}\left(x_{1}, x_{2}, x_{3}\right)=\left(x_{1} \wedge x_{2}\right) \vee\left(x_{2} \wedge x_{3}\right) \vee\left(x_{3} \wedge x_{1}\right)
$$

## Proof.

$$
\begin{aligned}
\phi(\mathbf{x}) & =\operatorname{median}\left(\phi\left(0_{i}, \mathbf{x}\right), x_{i}, \phi\left(1_{i}, \mathbf{x}\right)\right) \\
& =\underbrace{\left(\phi\left(0_{i}, \mathbf{x}\right) \wedge x_{i}\right)}_{\text {redundant }} \vee\left(x_{i} \wedge \phi\left(1_{i}, \mathbf{x}\right)\right) \vee \underbrace{\left(\phi\left(1_{i}, \mathbf{x}\right) \wedge \phi\left(0_{i}, \mathbf{x}\right)\right)}_{=\phi\left(0_{i}, \mathbf{x}\right)} \\
& =x_{i} \phi\left(1_{i}, \mathbf{x}\right) \amalg \phi\left(0_{i}, \mathbf{x}\right) \\
& =x_{i} \phi\left(1_{i}, \mathbf{x}\right)+\phi\left(0_{i}, \mathbf{x}\right)-x_{i} \underbrace{\phi\left(1_{i}, \mathbf{x}\right)}_{\text {redundant }} \phi\left(0_{i}, \mathbf{x}\right) \\
& =x_{i} \phi\left(1_{i}, \mathbf{x}\right)+\left(1-x_{i}\right) \phi\left(0_{i}, \mathbf{x}\right)
\end{aligned}
$$

## Reliability analysis

Reliability function of component $i \in[n]$

$$
R_{i}(t)=\operatorname{Pr}\left(T_{i}>t\right)=\operatorname{Pr}\left(X_{i}(t)=1\right)=\mathrm{E}\left[X_{i}(t)\right]
$$

$=$ probability that component $i$ does not fail in the interval $[0, t]$

System reliability function

$$
R_{S}(t)=\operatorname{Pr}\left(T_{S}>t\right)=\operatorname{Pr}\left(X_{S}(t)=1\right)=\mathrm{E}\left[X_{S}(t)\right]
$$

$=$ probability that the system does not fail in the interval $[0, t]$

## Reliability analysis

## Theorem. (Dukhovny 2007)

$$
R_{S}(t)=\sum_{A \subseteq[n]} v(A) \operatorname{Pr}\left(\mathbf{X}(t)=\mathbf{e}_{A}\right)
$$

Remarks.
(i) All the needed information is the distribution of $\mathbf{X}(t)$ (the knowledge of the joint distribution of $\mathbf{T}$ is not necessary)
(ii) The distribution of $\mathbf{X}(t)$ can be easily expressed in terms of the joint probability generating function of $\mathbf{X}(t)$

$$
G(\mathbf{z}, t)=\mathrm{E}\left[\prod_{i=1}^{n} z_{i}^{x_{i}(t)}\right] \quad\left(\left|z_{i}\right| \leqslant 1\right)
$$

We have

$$
\operatorname{Pr}\left(\mathbf{X}(t)=\mathbf{e}_{A}\right)=\sum_{B \subseteq A}(-1)^{|A|-|B|} G\left(\mathbf{e}_{B}, t\right)
$$

## Reliability analysis

When $T_{1}, \ldots, T_{n}$ are independent, we have

$$
R_{S}(t)=\sum_{A \subseteq[n]} v(A) \prod_{i \in A} R_{i}(t) \prod_{i \in[n] \backslash A}\left(1-R_{i}(t)\right)
$$

Alternative expression (Dukhovny and M. 2008)

$$
R_{S}(t)=\sum_{A \subseteq[n]} m_{v}(A) \operatorname{Pr}\left(T_{i}>t \forall i \in A\right)
$$

In case of independence

$$
R_{S}(t)=\sum_{A \subseteq[n]} m_{v}(A) \prod_{i \in A} R_{i}(t)
$$

## Mean time-to-failure of the system

The mean time-to-failure of the system is defined as

$$
\mathrm{MTTF}_{S}=\mathrm{E}\left[T_{S}\right]
$$

It is easy to show that

$$
\operatorname{MTTF}_{S}=\int_{0}^{\infty} R_{S}(t) d t
$$

(Rausand and Høyland 2004)

In case of independence

$$
\begin{aligned}
& \operatorname{MTTF}_{S}=\sum_{A \subseteq[n]} v(A) \int_{0}^{\infty} \prod_{i \in A} R_{i}(t) \prod_{i \in[n] \backslash A}\left(1-R_{i}(t)\right) d t \\
& \operatorname{MTTF}_{S}=\sum_{A \subseteq[n]} m_{v}(A) \int_{0}^{\infty} \prod_{i \in A} R_{i}(t) d t
\end{aligned}
$$

## Mean time-to-failure of the system

Example. Assume $R_{i}(t)=e^{-\lambda_{i} t}, i=1, \ldots, n$

$$
\begin{aligned}
\operatorname{MTTF}_{S} & =\sum_{A \subseteq[n]} m_{v}(A) \int_{0}^{\infty} \prod_{i \in A} e^{-\lambda_{i} t} d t \\
& =\sum_{A \subseteq[n]} m_{v}(A) \int_{0}^{\infty} e^{-\lambda_{A} t} d t \quad\left(\lambda_{A}=\sum_{i \in A} \lambda_{i}\right) \\
& =\sum_{\substack{A \subseteq[n] \\
A \neq \varnothing}} m_{v}(A) \frac{1}{\lambda_{A}}
\end{aligned}
$$

Series structure: $\quad$ MTTF $_{S}=\frac{1}{\lambda_{[n]}}$
Parallel structure: MTTF $_{S}=\sum_{\substack{A \subseteq[n] \\ A \neq \varnothing}}(-1)^{|A|-1} \frac{1}{\lambda_{A}}$

## Another advantage of the lattice polynomial language

## Generalization to weighted lattice polynomial functions

Suppose there are
(i) collective upper bounds on lifetimes of certain subsets of units (imposed by the physical properties of the assembly)

subset lifetime $=T \wedge c$
(ii) collective lower bounds (imposed by back-up blocks with constant lifetimes)

subset lifetime $=T \vee c$

## Another advantage of the lattice polynomial language

The lifetime of a general system with upper and/or lower bounds can be described through a weighted lattice polynomial function

$$
T_{S}=p\left(T_{1}, \ldots, T_{n}\right)
$$

## Example.



Suppose that the lifetime of component \#2 must lies in the time interval $[c, d]$

$$
\begin{aligned}
T_{S} & =T_{1} \wedge \operatorname{median}\left(c, T_{2}, d\right) \\
& =T_{1} \wedge\left(c \vee\left(T_{2} \wedge d\right)\right) \\
& =\left(c \wedge T_{1}\right) \vee\left(d \wedge T_{1} \wedge T_{2}\right)
\end{aligned}
$$

## Weighted lattice polynomial functions

The class of $n$-ary weighted lattice polynomial (w.l.p.) functions is defined as follows:
(i) For any $k \in[n]$ and any $c \in L$, the projection $\left(t_{1}, \ldots, t_{n}\right) \mapsto t_{k}$ and the constant function $\left(t_{1}, \ldots, t_{n}\right) \mapsto c$ are $n$-ary w.l.p. function
(ii) If $p$ and $q$ are $n$-ary w.l.p. functions then $p \wedge q$ and $p \vee q$ are $n$-ary w.l.p. functions
(iii) Every $n$-ary w.l.p. function is constructed by finitely many applications of the rules (i) and (ii).

## Weighted lattice polynomial functions

$$
\begin{array}{ll}
\begin{array}{ll}
\text { w.l.p. function } \longleftrightarrow & \text { set function } \\
p: L^{n} \rightarrow L & w: 2^{[n]} \rightarrow L \\
& w(A)=p\left(\mathbf{e}_{A}^{a, b}\right) \\
& A \subseteq[n]
\end{array}
\end{array}
$$

$\rightarrow$ We write $p_{w}$ instead of $p$

Representations of a w.l.p. function (Goodstein 1967)

$$
p_{w}(\mathbf{t})=\bigvee_{A \subseteq[n]}\left(w(A) \wedge \bigwedge_{i \in A} t_{i}\right)
$$

## Weighted lattice polynomial functions

Example (cont'd)

$$
\begin{gathered}
\bullet 1 \\
T_{S}=\left(c \wedge T_{1}\right) \vee\left(d \wedge T_{1} \wedge T_{2}\right) \\
p_{w}\left(t_{1}, t_{2}\right)=\left(c \wedge t_{1}\right) \vee\left(d \wedge t_{1} \wedge t_{2}\right) \\
\begin{array}{c|c|}
\hline A & w(A) \\
\hline \varnothing & 0 \\
\{1\} & c \\
\{2\} & 0 \\
\{1,2\} & d \\
\hline
\end{array}
\end{gathered}
$$

We can show that

$$
X_{S}(t)=\left(\operatorname{Ind}(c>t) X_{1}(t)\right) \amalg\left(\operatorname{Ind}(d>t) X_{1}(t) X_{2}(t)\right)
$$

## Weighted lattice polynomial functions

Representation of w.l.p. functions (DNF)

$$
p_{w}(\mathbf{t})=\bigvee_{A \subseteq[n]} w(A) \wedge \bigwedge_{i \in A} t_{i}
$$

Theorem. (Dukhovny and M. 2008)
If $T_{S}=p_{w}\left(T_{1}, \ldots, T_{n}\right)$ then

$$
X_{S}(t)=\coprod_{A \subseteq[n]} v_{t}(A) \prod_{i \in A} X_{i}(t) \quad(t \geqslant 0)
$$

where $v_{t}(A)=\operatorname{Ind}(w(A)>t)$

## Reliability analysis

Exact reliability formulas (Dukhovny and M. 2008)

$$
\begin{aligned}
& R_{S}(t)=\sum_{A \subseteq[n]} v_{t}(A) \operatorname{Pr}\left(\mathbf{X}(t)=\mathbf{e}_{A}\right) \\
& R_{S}(t)=\sum_{A \subseteq[n]} m_{v_{t}}(A) \operatorname{Pr}\left(T_{i}>t \forall i \in A\right)
\end{aligned}
$$

In case of independence

$$
\begin{aligned}
& R_{S}(t)=\sum_{A \subseteq[n]} v_{t}(A) \prod_{i \in A} R_{i}(t) \prod_{i \in[n] \backslash A}\left(1-R_{i}(t)\right) \\
& R_{S}(t)=\sum_{A \subseteq[n]} m_{v_{t}}(A) \prod_{i \in A} R_{i}(t)
\end{aligned}
$$

## Mean time-to-failure of the system

$$
\begin{aligned}
\operatorname{MTTF}_{S} & =\int_{0}^{\infty} R_{S}(t) d t \\
& =\sum_{A \subseteq[n]} \int_{0}^{\infty} m_{v_{t}}(A) \prod_{i \in A} R_{i}(t) d t \\
& =\sum_{A \subseteq[n]} \int_{0}^{\infty}\left(\sum_{B \subseteq A}(-1)^{|A|-|B|} v_{t}(B)\right) \prod_{i \in A} R_{i}(t) d t \\
& =\sum_{A \subseteq[n]} \sum_{B \subseteq A}(-1)^{|A|-|B|} \int_{0}^{\infty} \operatorname{Ind}(w(B)>t) \prod_{i \in A} R_{i}(t) d t \\
& =\sum_{A \subseteq[n]} \sum_{B \subseteq A}(-1)^{|A|-|B|} \int_{0}^{w(B)} \prod_{i \in A} R_{i}(t) d t
\end{aligned}
$$

## Mean time-to-failure of the system

Example. Assume $R_{i}(t)=e^{-\lambda_{i} t}, i=1, \ldots, n$

$$
\begin{aligned}
\operatorname{MTTF}_{S} & =\sum_{A \subseteq[n]} \sum_{B \subseteq A}(-1)^{|A|-|B|} \int_{0}^{w(B)} \prod_{i \in A} e^{-\lambda_{i} t} d t \\
& =\sum_{A \subseteq[n]} \sum_{B \subseteq A}(-1)^{|A|-|B|} \int_{0}^{w(B)} e^{-\lambda_{A} t} d t \quad\left(\lambda_{A}=\sum_{i \in A} \lambda_{i}\right) \\
& =w(\varnothing)+\sum_{\substack{A \subseteq[n] \\
A \neq \varnothing}} \sum_{B \subseteq A}(-1)^{|A|-|B|} \frac{1-e^{-\lambda_{A} w(B)}}{\lambda_{A}}
\end{aligned}
$$

## Conclusion

We have discussed the formal parallelism between two representations of systems

- Structure functions
- Lattice polynomial functions
$\rightarrow$ Their languages are equivalent in many ways


## Advantages

- Generalization to w.l.p. functions + exact reliability formulas
- Exact formulas for the distribution functions of w.l.p. functions of random variables

$$
Y=p_{w}\left(X_{1}, \ldots, X_{n}\right)
$$

- Several special cases can be investigated
- Symmetric w.l.p. functions : $w(A)=f(|A|)$
- The reliability of any subsystem depends only on the number of units in the subsystem $\operatorname{Pr}\left(\mathbf{X}(t)=\mathbf{e}_{A}\right)=g_{t}(|A|)$
- ...


## Thank you for your attention !

A. Dukhovny and J.-L. Marichal, Reliability analysis of semicoherent systems through their lattice polynomial descriptions
arXiv : 0809.1332

