# POLITECNICO DI TORINO Repository ISTITUZIONALE

Comparison results for inactivity times of k-out-of-n and general coherent systems with dependent components

Original

Comparison results for inactivity times of k-out-of-n and general coherent systems with dependent components / Navarro, Jorge; Longobardi, Maria; Pellerey, Franco. - In: TEST. - ISSN 1133-0686. - STAMPA. - 26:4(2017), pp. 822-846. [10.1007/s11749-017-0535-5]

*Availability:* This version is available at: 11583/2688420 since: 2017-10-30T18:36:34Z

Publisher: Springer New York LLC

Published DOI:10.1007/s11749-017-0535-5

Terms of use:

This article is made available under terms and conditions as specified in the corresponding bibliographic description in the repository

Publisher copyright Springer postprint/Author's Accepted Manuscript

This version of the article has been accepted for publication, after peer review (when applicable) and is subject to Springer Nature's AM terms of use, but is not the Version of Record and does not reflect post-acceptance improvements, or any corrections. The Version of Record is available online at: http://dx.doi.org/10.1007/s11749-017-0535-5

(Article begins on next page)

Comparison results for inactivity times of k-out-of-n and general coherent systems with dependent components

Authors version Published in: TEST 26 (2017), pp 822–846 DOI 10.1007/s11749-017-0535-5, URL: https://link.springer.com/article/10.1007/s11749-017-0535-5

Jorge Navarro · Maria Longobardi · Franco Pellerey

October 19, 2017

Abstract Consider a coherent system with possibly dependent components having lifetime T, and assume we know that it failed before a given time t > 0. Its inactivity time t - T can be evaluated under different conditional events. In fact, one might just know that the system has failed and then consider the inactivity time  $(t - T|T \le t)$ , or one may also know which ones of the components have failed before time t, and then consider the corresponding system's inactivity time under this condition. For all these cases we obtain a representation of the reliability function of system inactivity time based on distortion functions, which, in turn, includes a description of the structure of dependence between components through the copula of the vector of components' lifetimes. Making use of these representations, new stochastic comparison results for inactivity times under the different conditional events are provided, as well as comparison results for inactivity times of systems having different structure functions. These results also apply to order statistics, being the order statistics particular cases of coherent systems (k-out-of-n systems).

Jorge Navarro Facultad de Matemáticas Universidad de Murcia 30100 Murcia, Spain E-mail: jorgenav@um.es

### Maria Longobardi

Dipartimento di Matematica e Applicazioni Università di Napoli FEDERICO II Via Cintia, Napoli, Italy E-mail: maria.longobardi@unina.it

#### Franco Pellerey

Dipartimento di Scienze Matematiche Politecnico di Torino C.so Duca degli Abruzzi 24 I-10129 Torino, Italy E-mail: franco.pellerey@polito.it **Keywords** Reliability  $\cdot$  inactivity time  $\cdot$  coherent systems  $\cdot$  stochastic orders  $\cdot$  distortion functions  $\cdot$  copula

AMS 2000 Subject Classification: Primary 62K10, 60E15 Secondary 90B25

## **1** Introduction

In reliability theory, analysis of coherent systems is a relevant topic since most of multi-component systems can be modeled through them (see, e.g., [1] and [10] for a detailed introduction to this subject, related properties and examples of application). Series systems, parallel systems, k-out-of-n systems (order statistics) are examples of coherent systems. In this field, it is important to study the performance of a system composed by different kinds of units, maybe having dependent lifetimes, in order to evaluate their reliability or to provide bounds for related quantities such as their failure rates or expected lifetimes. Some results on this topic were given, e.g., in [14, 20–22, 26] and the references therein.

In particular, special attention has been paid in the study of the system residual lifetime under different assumptions (see, e.g., [11, 12, 14, 15, 25]). Thus, for example, at a given time t > 0 we may just now that the system is working or we may have more information about the component states (all of them are working, some of them are working and some have failed before t, etc.). However, in some situations, the interest may be on the past lifetime of a system and not only on the future, i.e., on its inactivity time, having observed that the system is failed at a given time t (see [7,11,28, 29]).

Let T be the lifetime of the system, and let  $X_i$ , i = 1, ..., n, be the lifetimes of its components. Dealing with inactivity times, different conditions can be assumed observing that the system has failed at a time t > 0. In fact, one can just know that the system lifetime is smaller than t, i.e., T < t, or one can know, for example, that all its components have failed before t, i.e.,  $X_i < t$ ,  $\forall i = 1, ..., n$ . In this particular case, one can believe that the inactivity time in the first case is smaller, in some stochastic sense, than the inactivity time in the second case. That is, for example, one can affirm that the stochastic inequality

$$(t - T|T \le t) \le_{ST} (t - T|X_1 \le t, \dots, X_n \le t) \quad \forall t \ge 0,$$
 (1.1)

holds true for every coherent system (a formal definition of the stochastic comparison  $\leq_{ST}$  will be given in Section 3). However, as shown in the Example 4 (see Section 5), this assertion is not always satisfied.

Motivated by this example, this paper provides a study on the inactivity time of coherent systems formed by a number n of components with possibly dependent lifetimes, considering different conditioning events on the failed components in the system. For all of them we give new representations for the reliability functions of the corresponding inactivity times, and we apply them proving simple conditions for comparisons of inactivity times according to the most important stochastic orders considered in reliability theory.

The paper is organized as follows. Section 2 firstly recall the notion of distortion functions, which have been recently introduced in the literature and used to formally describe how the dependence structure between components affects the lifetime of a system (see [18,22,23]). Then, the representations of the reliability function of inactivity times of coherent systems based on distortion functions, under different conditioning, are provided, and some immediate consequences of these representations are described. Section 3 contains conditions to compare inactivity times under the different conditional events, as well as, comparison results for inactivity times of systems having different structure functions. The final Section 4 is devoted to some illustrative examples and counterexamples.

Throughout the paper, the notation (X|A) is used to represent a random variable whose distribution is the conditional distribution of X given the event A (assuming Pr(A) > 0). Also, whenever we consider a ratio  $\frac{a}{b}$ , we assume  $b \neq 0$  unless otherwise indicated. The notation g' represents the derivative of the function g and, whenever we write g', we assume that this derivative exists. Finally, the terms "increasing" and "decreasing" are used in non-strict sense.

#### 2 Representation of inactivity times through distortion functions

Some basic notions of coherent systems are provided now. Given a multicomponent system, its structure function  $\phi : \{0,1\}^n \to \{0,1\}$  is a function that maps the state vector  $(\hat{x}_1,\ldots,\hat{x}_n)$  of its n components (where  $\hat{x}_i = 1$  if component i is working and  $\hat{x}_i = 0$  if it is failed) to the state  $\hat{y} \in \{0,1\}$  of the system itself. The system is said to be *coherent* whenever every component is relevant (i.e., it affects the working or failure of the system) and the structure function is monotone in every component. Also, given a coherent system with n possibly dependent components having lifetimes  $X_1,\ldots,X_n \ge 0$ , the relationship between the vector  $(X_1,\ldots,X_n)$  of component's lifetimes and system's lifetime T is described by the relation  $T = \tau(X_1,\ldots,X_n)$ , where the *coherent life function*  $\tau : \mathbb{R}^n \to \mathbb{R}$  is defined as

$$\tau(x_1,\ldots,x_n) = \sup\{t \ge 0 : \phi(\widehat{x}_{1,t},\ldots,\widehat{x}_{n,t}) = 1\},\$$

where  $\widehat{x}_{i,t} = 1$  if  $x_i > t$ , or  $\widehat{x}_{i,t} = 0$  if  $x_i \le t$ , for  $i \in \{1, \ldots, n\}$ .

For the sequel it will be useful to recall that a subset  $C \subseteq \{1, \ldots, n\}$  of the components indices is said to be a *cut set* if the system does not work whenever the components indexed in C do not work. The set is a *minimal cut set* if it is a minimal set of elements whose failure causes the system to fail. Similarly, a subset  $\mathcal{P} \subseteq \{1, \ldots, n\}$  is a *path set* if the system works whenever the components indexed in  $\mathcal{P}$  work, and it is called *minimal path set* if it does not contain other path sets. We refer the reader to [1] for further details on coherent systems.

We now recall the concept of copula of a random vector, which is needed for the representation of the distribution of inactivity times of systems through distortion functions. First, recall that, for every dimension  $n \ge 2$  a *copula* is an *n*-dimensional distribution function concentrated on  $[0,1]^n$ whose univariate marginals are uniformly distributed on  $[0,1] \subseteq \mathbb{R}$  (see the monographs [5] or [24] for details). Let  $(X_1, \ldots, X_n)$  be a random vector with joint distribution function F and marginal distribution functions  $F_i, i \in \{1, \ldots, n\}$ . Then the joint distribution F can be represented as

$$F(x_1,\ldots,x_n) = C(F_1(x_1),\ldots,F_n(x_n))$$

for a copula C. Notice that, as affirmed by the well-known Sklar's theorem, if the marginal distribution functions  $F_i$  are continuous, then the copula C of the vector  $(X_1, \ldots, X_n)$  is unique and it is given by

$$C(u_1,\ldots,u_n) = F(F_1^{-1}(u_1),\ldots,F_n^{-1}(u_n)),$$

for all  $u_i \in [0, 1], i \in \{1, ..., n\}$ , where the  $F_i^{-1}$  are the pseudo-inverses of the  $F_i$ . We will assume here, and everywhere throughout the paper, such a continuity property.

In a similar way the joint reliability function  $\overline{F}$  can be represented as

$$\overline{F}(x_1,\ldots,x_n) = \overline{C}(\overline{F}_1(x_1),\ldots,\overline{F}_n(x_n))$$

where  $\overline{F}_i, i \in \{1, \ldots, n\}$  are the marginal reliability functions and  $\overline{C}$  is a copula called *survival copula* of  $(X_1, \ldots, X_n)$ . Similarly as above,

$$\overline{C}(u_1,\ldots,u_n)=\overline{F}(\overline{F}_1^{-1}(u_1),\ldots,\overline{F}_n^{-1}(u_n)),$$

for all  $u_i \in [0, 1], i \in \{1, \dots, n\}.$ 

Let now T be the lifetime of a coherent system with structure function  $\phi$  and with n possibly dependent components having lifetimes  $X_1, \ldots, X_n \ge 0$ . Denote with F the joint distribution function of the vector of components' lifetimes, with C its copula, and with  $F_i$  the distribution function of  $X_i$ ,  $i = 1, \ldots, n$ . Analogously, let  $\overline{F}$  denotes the joint reliability function of  $(X_1, \ldots, X_n)$ , with  $\overline{C}$  its survival copula, and with  $\overline{F}_i$  the reliability functions of the component's lifetimes. Then a representation of the distribution of T similar to the above copula representations was obtained in [23] (see also [17,22]). According to such a representation, the system reliability  $\overline{F}_T(t) = \Pr(T > t)$ can be written as

$$\overline{F}_T(t) = \overline{Q}(\overline{F}_1(t), \dots, \overline{F}_n(t)), \qquad (2.1)$$

where  $\overline{Q}$ :  $[0,1]^n \to [0,1]$  is a continuous increasing function satisfying  $\overline{Q}(0,\ldots,0) = 0$  and  $\overline{Q}(1,\ldots,1) = 1$ , which only depends on the system structure  $\phi$  and on the survival copula  $\overline{C}$  of the vector  $(X_1,\ldots,X_n)$ . In other words,  $\overline{Q}$  is simply a continuous aggregation function (for definition and examples of aggregation functions see, e.g., [4,8]). It should be pointed out that the function  $\overline{Q}$  is not necessarily a copula. In fact,  $\overline{Q}$  can be expressed in terms of the survival copula  $\overline{C}$ as follows. Assume that the system admits a number r of minimal path sets  $\mathcal{P}_1, \ldots, \mathcal{P}_r$ , and denote  $I_r = \{1, \ldots, r\}$ . Then

$$\overline{Q}(u_1,\ldots,u_n) = \sum_{\emptyset \neq I \subseteq I_r} (-1)^{|I|+1} \overline{C}_I(u_1,\ldots,u_n)$$
(2.2)

where |I| is the cardinality of the set I,  $\overline{C}_I(u_1, \ldots, u_n) = \overline{C}(\widetilde{u}_1^I, \ldots, \widetilde{u}_n^I)$  and  $\widetilde{u}_k^I = u_k$  whenever  $k \in \bigcup_{m \in I} \mathcal{P}_m$ , or  $\widetilde{u}_k^I = 1$  whenever  $k \notin \bigcup_{m \in I} \mathcal{P}_m$ . A similar representation holds for the respective distribution function:

$$F_T(t) = Q(F_1(t), \dots, F_n(t)),$$
 (2.3)

where, similarly as above, assuming that the system admits minimal cut sets  $C_1, \ldots, C_s$ , it holds

$$Q(u_1, \dots, u_n) = \sum_{\emptyset \neq I \subseteq I_s} (-1)^{|I|+1} C_I(u_1, \dots, u_n)$$
(2.4)

where  $I_s = \{1, \ldots, s\}$ ,  $C_I(u_1, \ldots, u_n) = C(\tilde{u}_1^I, \ldots, \tilde{u}_n^I)$  and  $\tilde{u}_k^I = u_k$  whenever  $k \in \bigcup_{i \in I} C_i$ , or  $\tilde{u}_k^I = 1$ whenever  $k \notin \bigcup_{i \in I} C_i$ . In the particular case that the  $X_i$  are independent, then the previous expression for Q reduces to

$$Q(u_1,\ldots,u_n) = \sum_{\emptyset \neq I \subseteq I_s} (-1)^{|I|+1} \prod_{k \in \bigcup_{i \in I} \mathcal{C}_i} u_k.$$

$$(2.5)$$

It should be observed that

$$Q(u_1,\ldots,u_n) = 1 - \overline{Q}(1-u_1,\ldots,1-u_n)$$

for all  $(u_1, \ldots, u_n) \in [0, 1]^n$ . Representations (2.1) and (2.3) are equivalent but sometimes it is better to work with (2.1) instead of (2.3) (and vice versa). When the components are identically distributed, that is,  $F_1 = \cdots = F_n$ , these representations can be reduced to

$$\overline{F}_T(t) = \overline{q}(\overline{F}_1(t)) \tag{2.6}$$

and

$$F_T(t) = q(F_1(t)),$$
 (2.7)

where  $\overline{q}(u) = \overline{Q}(u, \ldots, u)$  and  $q(u) = Q(u, \ldots, u) = 1 - \overline{q}(1-u)$ ,  $u \in [0, 1]$ . The distributions that can be written as in (2.6) and (2.7) are called *distorted distribution* and the functions q and  $\overline{q}$  are called respectively *distortion and dual distortion functions* (see, e.g., [16] and the references therein). The distributions that can be written as in (2.1) and (2.3) are called *generalized distorted distributions* (see [17,18,20]). The functions Q and  $\overline{Q}$  are called *generalized distortion functions*.

In particular, for the series system with n components, we have  $T = X_{1:n} = \min(X_1, \ldots, X_n)$ and

$$\overline{F}_{1:n}(t) = \overline{C}(\overline{F}_1(t), \dots, \overline{F}_n(t)),$$

that is,  $\overline{Q}_{1:n} = \overline{C}$  (and it is obviously a copula). If the components are identically distributed, then  $\overline{q}_{1:n}$  is the diagonal section of  $\overline{C}$  (i.e., the function  $\delta : [0,1] \to [0,1]$  defined as  $\delta(u) = \overline{C}(u, \ldots, u)$  for all  $u \in [0,1]$ ). Analogously, for the parallel system with n components, we have  $T = X_{n:n} = \max(X_1, \ldots, X_n)$  and

$$F_{n:n}(t) = C(F_1(t), \dots, F_n(t))$$

that is,  $Q_{n:n} = C$ . If the components are identically distributed, then  $q_{n:n}$  is the diagonal section of C.

Now, we provide similar representations for the distributions of inactivity times of the system, that is, the time without service  $(t - T|A_t)$  under different assumptions  $A_t$  which imply  $T \leq t$ . In fact, assuming that a coherent system starts to work at time 0 and it is failed at time t > 0, we might have different information about the states of the components. We can thus consider the following reasonable cases.

**Case 1:** The less informative case is to consider that we only know that the system has failed at time t. Then it is easy to observe that the system inactivity time is

$$T_t = (t - T | T \le t).$$

Its reliability function is obtained in the following proposition. Before we need to note that if  $F_i(t) > 0$ , then the reliability function  $\overline{F}_{i,t}$  of the *i*th component inactivity time  $(t - X_i | X_i \le t)$  is given by

$$\overline{F}_{i,t}(x) = \Pr(t - X_i > x | X_i \le t) = \frac{F_i(t - x)}{F_i(t)}$$

$$(2.8)$$

for  $x \in [0, t]$  and i = 1, ..., n. These reliability functions will be used to represent the reliability function of the system inactivity time.

**Proposition 1** If  $F_i(t) > 0$  for i = 1, ..., n, then the reliability function of  $T_t$  can be written as

$$\overline{F}_t(x) = \overline{Q}_t(\overline{F}_{1,t}(x), \dots, \overline{F}_{n,t}(x))$$
(2.9)

for  $x \in [0, t]$ , where

$$\overline{Q}_t(u_1,\ldots,u_n) = \frac{Q(u_1F_1(t),\ldots,u_nF_n(t))}{Q(F_1(t),\ldots,F_n(t))}$$

is a generalized distortion function which depends on the distortion function Q defined in (2.4) and on the values  $F_i(t)$ , i = 1, ..., n.

*Proof* For  $x \in [0, t]$ , from (2.3), we have

$$\overline{F}_t(x) = \Pr(t - T > x | T \le t) = \frac{\Pr(T < t - x)}{\Pr(T \le t)}$$
$$= \frac{F_T(t - x)}{F_T(t)} = \frac{Q(F_1(t - x), \dots, F_n(t - x))}{Q(F_1(t), \dots, F_n(t))}$$
$$= \frac{Q(F_1(t)\overline{F}_{1,t}(x), \dots, F_n(t)\overline{F}_{n,t}(x))}{Q(F_1(t), \dots, F_n(t))} = \overline{Q}_t(\overline{F}_{1,t}(x), \dots, \overline{F}_{n,t}(x))$$

which finishes the proof.

**Case 2:** Here we assume that we know the set  $W \subseteq \{1, \ldots, n\}$  of indices of components that are working at time t (and so the set  $W^c = \{1, \ldots, n\} - W$  of those that have failed), that is,  $A_t = \{X_W > t, X^{W^c} \le t\}$ , where  $X_W = \min_{i \in W} X_i$  (lifetime of the series system with components W),  $X^{W^c} = \max_{i \in W^c} X_i$  (lifetime of the parallel system with components  $W^c$ ). Of course, this assumption implies that the components may work even if the system has failed, and that  $\{X^{W^c} \le t\}$ implies  $\{T \le t\}$  (i.e.  $W^c$  is a cut set). Also  $W \ne \{1, \ldots, n\}$ . Then we can consider the following system inactivity time

$$T_t^W = (t - T | X_W > t, X^{W^c} \le t).$$

Note that here we include the particular case in which all the components have failed at time t, that is,  $W = \emptyset$  and  $W^c = \{1, \ldots, n\}$ . We obtain a representation similar to (2.9) for  $T_t^W$  in the following proposition.

**Proposition 2** If  $F_i(t) > 0$  for i = 1, ..., n, then the reliability function of  $T_t^W$  can be written as

$$\overline{F}_{t}^{W}(x) = \overline{Q}_{t}^{W}(\overline{F}_{1,t}(x), \dots, \overline{F}_{n,t}(x))$$
(2.10)

for  $x \in [0, t]$ , where the reliability functions  $\overline{F}_{i,t}(x)$  are defined as in (2.8) and  $\overline{Q}_t^W$  is a generalized distortion function. If  $C_1, \ldots, C_s$  are the minimal path sets of the system, then  $\overline{Q}_t^W$  is given by

$$\overline{Q}_{t}^{W}(u_{1},\ldots,u_{n}) = \frac{\sum_{\emptyset \neq I \subseteq I_{s}} \sum_{A \subseteq W} (-1)^{|I|+|A|+1} C_{I,A,W}(u_{1},\ldots,u_{n})}{\sum_{A \subseteq W} (-1)^{|A|} C_{A,W}(F_{1}(t),\ldots,F_{n}(t))},$$
(2.11)

where  $C_{I,A,W}(u_1,\ldots,u_n) = 0$  when  $W \cap \bigcup_{i \in I} \mathcal{C}_i \neq \emptyset$  or

$$C_{I,A,W}(u_1,\ldots,u_n) = C(\widetilde{u}_1^{I,A,W},\ldots,\widetilde{u}_n^{I,A,W})$$

when  $\cup_{i \in I} \mathcal{C}_i \subseteq W^c$ , where  $\widetilde{u}_k^{I,A,W} = F_k(t)$  whenever  $k \in A \cup (W^c - \cup_{i \in I} \mathcal{C}_i)$ , or  $\widetilde{u}_k^{I,A,W} = 1$  whenever  $k \in W - A$ , or  $\widetilde{u}_k^{I,A,W} = u_k F_k(t)$  whenever  $k \in \bigcup_{i \in I} \mathcal{C}_i$ , and where

$$C_{A,W}(u_1,\ldots,u_n)=C(\widetilde{u}_1^{A,W},\ldots,\widetilde{u}_n^{A,W})$$

and  $\widetilde{u}_k^{A,W} = u_k$  whenever  $k \in A \cup W^c$ , or  $\widetilde{u}_k^{A,W} = 1$  whenever  $k \in W - A$ .

*Proof* From the definition, we have

$$\overline{F}_t^W(x) = \Pr(t - T > x | X_W > t, X^{W^c} \le t)$$
$$= \frac{\Pr(T < t - x, X_W > t, X^{W^c} \le t)}{\Pr(X_W > t, X^{W^c} \le t)}.$$

If  $C_1, \ldots, C_r$  are the minimal cut sets and denoting again  $X^{\mathcal{C}} = \max_{i \in \mathcal{C}} X_i$ , one has

$$\overline{F}_{t}^{W}(x) = \frac{\Pr(T < t - x, X_{W} > t, X^{W^{c}} \le t)}{\Pr(X_{W} > t, X^{W^{c}} \le t)}$$

$$= \frac{\Pr(\min_{j=1,\dots,s} X^{\mathcal{C}_{j}} < t - x, X_{W} > t, X^{W^{c}} \le t)}{\Pr(X_{W} > t, X^{W^{c}} \le t)}$$

$$= \frac{\Pr(\cup_{j=1,\dots,s} \{X^{\mathcal{C}_{j}} < t - x\}, X_{W} > t, X^{W^{c}} \le t)}{\Pr(X_{W} > t, X^{W^{c}} \le t)}.$$

The denominator in the preceding expression can be written in terms of C and  $F_1, \ldots, F_n$ , as

$$D = \Pr(X_W > t, X^{W^c} \le t) = \sum_{A \subseteq W} (-1)^{|A|} C_{A,W}(F_1(t), \dots, F_n(t))$$

where  $C_{A,W}$  is defined in the statement.

A similar representation holds for the numerator

$$N = \Pr(\bigcup_{i=1,\dots,s} \{ X^{\mathcal{C}_i} < t - x \}, X_W > t, X^{W^c} \le t )$$
  
=  $\sum_{\emptyset \neq I \subseteq I_s} (-1)^{|I|+1} \Pr(X^{\bigcup_{i \in I} \mathcal{C}_i} < t - x, X_W > t, X^{W^c} \le t )$   
=  $\sum_{\emptyset \neq I \subseteq I_s} \sum_{A \subseteq W} (-1)^{|I|+|A|+1} C_{I,A,W}(\overline{F}_{1,t}(x),\dots,\overline{F}_{1,t}(x))$ 

where  $C_{I,A,W}$  is defined in the statement. Therefore, the final expression for  $\overline{Q}_t^W$  is obtained by using such expressions for N and D.

Note that  $\overline{Q}_t^W$  only depends on  $u_i$  for  $i \in W^c$  (i.e., it is constant in  $u_i$  for  $i \in W$ ). As a particular case, whenever  $W = \emptyset$ , then (2.11) reduces to

$$Q_t^{\emptyset}(u_1, \dots, u_n) = \frac{\sum_{\emptyset \neq I \subseteq I_s} (-1)^{|I|+1} C_{I,\emptyset,\emptyset}(u_1, \dots, u_n)}{C(F_1(t), \dots, F_n(t))}$$
(2.12)

where  $C_{I,\emptyset,\emptyset}(u_1,\ldots,u_n) = C(\widetilde{u}_1^{I,\emptyset,\emptyset},\ldots,\widetilde{u}_n^{I,\emptyset,\emptyset})$  and where  $\widetilde{u}_k^{I,\emptyset,\emptyset} = F_k(t)$  whenever  $k \notin \bigcup_{i \in I} \mathcal{C}_i$ , or  $\widetilde{u}_k^{I,\emptyset,\emptyset} = u_k F_k(t)$  whenever  $k \in \bigcup_{i \in I} \mathcal{C}_i$ .

Let us see now two examples showing how these representations can be obtained.

Example 1 The simplest case of application of the above representations is in a series system with two possibly dependent components, i.e., with lifetime  $T = \min(X_1, X_2)$ . Its reliability function is

$$\overline{F}_T(t) = \overline{C}(\overline{F}_1(t), \overline{F}_2(t))$$

and its distribution function is

$$F_T(t) = \Pr(\min(X_1, X_2) \le t) = F_1(t) + F_2(t) - C(F_1(t), F_2(t)) = Q(F_1(t), F_2(t))$$

where  $F_1, F_2$  are the components' continuous distribution functions and

$$Q(u_1, u_2) = u_1 + u_2 - C(u_1, u_2) = 1 - \overline{C}(1 - u_1, 1 - u_2).$$

In case of independence between lifetimes' components, clearly  $Q(u_1, u_2) = Q_{\perp}(u_1, u_2) = u_1 + u_2 - u_1 u_2$ .

If at time t > 0 we just know that the system has failed, that is,  $T \le t$ , then the reliability function of  $(t - T | T \le t)$  is

$$\overline{F}_t(x) = \frac{F_T(t-x)}{F_T(t)} = \overline{Q}_t(\overline{F}_{1,t}(x), \overline{F}_{2,t}(x))$$

for  $x \in [0, t]$ , where

$$\overline{Q}_t(u_1, u_2) = \frac{Q(u_1F_1(t), u_2F_2(t))}{Q(F_1(t), F_2(t))} = \frac{u_1F_1(t) + u_2F_2(t) - C(u_1F_1(t), u_2F_2(t))}{F_1(t) + F_2(t) - C(F_1(t), F_2(t))}$$

is a generalized distortion function. In particular, if the components are independent, then

$$\overline{Q}_t(u_1, u_2) = \frac{u_1 F_1(t) + u_2 F_2(t) - u_1 u_2 F_1(t) F_2(t)}{F_1(t) + F_2(t) - F_1(t) F_2(t)}$$

Another option is to assume that at time t > 0 we know that the first component is working and the second has failed, that is,  $W = \{1\}$ . Then the series system has failed,  $T \le t$ , and the reliability function of  $T_t^{\{1\}} = (t - T | X_1 > t, X_2 \le t)$  is

$$\overline{F}_t^{\{1\}}(x) = \frac{\Pr(T < t - x, X_1 > t, X_2 \le t)}{\Pr(X_1 > t, X_2 \le t)}, \quad x \in [0, t],$$

where

 $\Pr(T < t - x, X_1 > t, X_2 \le t) = \Pr(X_2 < t - x, X_1 > t) = \Pr(X_2 < t - x) - \Pr(X_1 \le t, X_2 < t - x)$ 

and

$$\Pr(X_1 > t, X_2 \le t) = \Pr(X_2 \le t) - \Pr(X_1 \le t, X_2 \le t).$$

Then

$$\overline{F}_{t}^{\{1\}}(x) = \overline{Q}_{t}^{\{1\}}(\overline{F}_{1,t}(x), \overline{F}_{2,t}(x))$$
(2.13)

where

$$\overline{Q}_t^{\{1\}}(u_1, u_2) = \frac{u_2 F_2(t) - C(F_1(t), u_2 F_2(t))}{F_2(t) - C(F_1(t), F_2(t))}$$

is a generalized distortion function. Note that  $\overline{Q}_t^{\{1\}}$  only depends on  $u_2$ . In particular, if the components are independent, then

$$\overline{Q}_t^{\{1\}}(u_1, u_2) = \frac{u_2 - u_2 F_1(t)}{1 - F_1(t)} = u_2.$$

that is,  $(t - T | X_W > t, X^{W^c} \le t)$  has the same law as  $(t - T^* | X^{W^c} \le t)$ , where  $T^*$  is the lifetime of the system obtained from the original one by deleting the cut sets which have at least an element in W (i.e.,  $T^* = \min_{\{j:C_j \cap W = \emptyset\}} X^{C_j}$ ,  $T^* = X_2$  in this example), as one can expect. The representation for the case in which the first component has failed and the second is working can be obtained in a similar way.

In this example we can also consider the case  $W = \emptyset$  (note that we cannot consider  $W = \{1, 2\}$ since  $X_W > t$  implies T > t). The reliability function of  $T_t^{\emptyset} = (t - T | X_1 < t, X_2 < t)$  is, for  $x \in [0, t]$ ,

$$\begin{split} \overline{F}_{t}^{\emptyset}(x) &= \frac{\Pr(T < t - x, X_{1} < t, X_{2} < t)}{\Pr(X_{1} < t, X_{2} < t)} \\ &= \frac{\Pr(X_{1} < t - x, X_{2} < t) + \Pr(X_{1} < t, X_{2} < t - x) - \Pr(X_{1} < t - x, X_{2} < t - x)}{\Pr(X_{1} < t, X_{2} < t)} \\ &= \overline{Q}_{t}^{\emptyset}((\overline{F}_{1,t}(x), \overline{F}_{2,t}(x)), \end{split}$$

where

$$\overline{Q}_t^{\emptyset}(u_1, u_2) = \frac{C(u_1 F_1(t), F_2(t)) + C(F_1(t), u_2 F_2(t)) - C(u_1 F_1(t), u_2 F_2(t))}{C(F_1(t), F_2(t))}$$

is a generalized distortion function. In particular, if the components are independent, then

$$\overline{Q}_t^{\emptyset}(u_1, u_2) = u_1 + u_2 - u_1 u_2 = Q_{\perp}(u_1, u_2),$$

where  $Q_{\perp}$  is the generalized distortion function of the series system in the case of independent components.

Example 2 Let us consider the system with lifetime  $T = \min(X_1, \max(X_2, X_3))$ . It may represent, for example, a server and two computers supporting the web page of a shop. The system works if the server works and, at least, a computer works. Its minimal cut sets are  $C_1 = \{1\}$  and  $C_2 = \{2, 3\}$ and its distribution function is

$$F_T(t) = \Pr(\min(X_1, \max(X_2, X_3)) \le t)$$
  
=  $F_1(t) + C(1, F_2(t), F_3(t)) - C(F_1(t), F_2(t), F_3(t))$   
=  $Q(F_1(t), F_2(t), F_3(t)),$ 

where  $Q(u_1, u_2, u_3) = u_1 + C(1, u_2, u_3) - C(u_1, u_2, u_3)$  and  $F_1, F_2, F_3$  are the continuous component distribution functions. Whenever the component's lifetimes are independent, then  $Q(u_1, u_2, u_3) = Q_{\perp}(u_1, u_2, u_3) = u_1 + u_2u_3 - u_1u_2u_3$ 

If at time t > 0, we just know that the system has failed, that is,  $T \le t$ , then the reliability function of  $(t - T|T \le t)$  is

$$\overline{F}_t(x) = \frac{F_T(t-x)}{F_T(t)} = \overline{Q}_t(\overline{F}_{1,t}(x), \overline{F}_{2,t}(x))$$

for  $x \in [0, t]$ , where the reliability functions  $\overline{F}_{i,t}(x)$  are defined as in (2.8) and

$$\begin{split} \overline{Q}_t(u_1, u_2, u_3) &= \frac{Q(u_1 F_1(t), u_2 F_2(t), u_3 F_3(t))}{Q(F_1(t), F_2(t), F_3(t))} \\ &= \frac{u_1 F_1(t) + C(1, u_2 F_2(t), u_3 F_3(t)) - C(u_1 F_1(t), u_2 F_2(t), u_3 F_3(t))}{F_1(t) + C(1, F_2(t), F_3(t)) - C(F_1(t), F_2(t), F_3(t))} \end{split}$$

is a generalized distortion function.

Another option is to assume that, at time t > 0, we know that all the components have failed, that is,  $W = \emptyset$ . Then the system has failed,  $T \leq t$ , and the reliability function of  $T_t^{\emptyset} = (t - T | X_1 \leq t, X_2 \leq t, X_3 \leq t)$  is

$$\overline{F}_t^{\emptyset}(x) = \frac{\Pr(T < t - x, X_1 \le t, X_2 \le t, X_3 \le t)}{\Pr(X_1 \le t, X_2 \le t, X_3 \le t)},$$

where

$$Pr(T < t - x, X_1 \le t, X_2 \le t, X_3 \le t) = Pr(X_1 < t - x, X_2 \le t, X_3 \le t) + Pr(X_1 \le t, X_2 < t - x, X_3 < t - x) - Pr(X_1 < t - x, X_2 < t - x, X_3 < t - x),$$

and

$$\Pr(X_1 \le t, X_2 \le t, X_3 \le t) = C(F_1(t), F_2(t), F_3(t)).$$

Then

$$\overline{F}_t^{\emptyset}(x) = \overline{Q}_t^{\emptyset}(\overline{F}_{1,t}(x), \overline{F}_{2,t}(x), \overline{F}_{3,t}(x)), \qquad (2.14)$$

where

$$\overline{Q}_{t}^{\emptyset}(u_{1}, u_{2}, u_{3}) = \frac{C(u_{1}F_{1}(t), F_{2}(t), F_{3}(t)) + C(F_{1}(t), u_{2}F_{2}(t), u_{3}F_{3}(t))}{C(F_{1}(t), F_{2}(t), F_{3}(t))} - \frac{C(u_{1}F_{1}(t), u_{2}F_{2}(t), u_{3}F_{3}(t))}{C(F_{1}(t), F_{2}(t), F_{3}(t))}$$

is a generalized distortion function. In particular, if the components are independent, then

$$\overline{Q}_t^{\emptyset}(u_1, u_2, u_3) = u_1 + u_2 u_3 - u_1 u_2 u_3 = Q_{\perp}(u_1, u_2, u_3),$$

where  $Q_{\perp}$  is the generalized distortion function of the system in the case of independent components.

Another option is to assume that at time t > 0, the only working component is the third component, that is,  $W = \{3\}$ . Then the system has failed,  $T \leq t$ , and the reliability function of  $T_t^{\{3\}} = (t - T | X_1 \leq t, X_2 \leq t, X_3 > t)$  is

$$\overline{F}_{t}^{\{3\}}(x) = \frac{\Pr(T < t - x, X_{1} \le t, X_{2} \le t, X_{3} > t)}{\Pr(X_{1} \le t, X_{2} \le t, X_{3} > t)},$$

where

$$Pr(T < t - x, X_1 \le t, X_2 \le t, X_3 > t) = Pr(X_3 > t, X_1 < t - x, X_2 \le t)$$
$$= Pr(X_1 < t - x, X_2 \le t) - Pr(X_1 < t - x, X_2 \le t, X_3 \le t)$$

and

$$\Pr(X_1 \le t, X_2 \le t, X_3 > t) = \Pr(X_1 \le t, X_2 \le t) - \Pr(X_3 \le t, X_1 \le t, X_2 \le t).$$

Then

$$\overline{F}_{t}^{\{3\}}(x) = \overline{Q}_{t}^{\{3\}}(\overline{F}_{1,t}(x), \overline{F}_{2,t}(x), \overline{F}_{3,t}(x)),$$
(2.15)

where

$$\overline{Q}_{t}^{\{3\}}(u_{1}, u_{2}, u_{3}) = \frac{C(u_{1}F_{1}(t), F_{2}(t), 1) - C(u_{1}F_{1}(t), F_{2}(t), F_{3}(t))}{C(F_{1}(t), F_{2}(t), 1) - C(F_{1}(t), F_{2}(t), F_{3}(t))}$$

is a generalized distortion function. Note that it only depends on  $u_1$ . In particular, if the components are independent, then

$$\overline{Q}_t^{\{3\}}(u_1, u_2, u_3) = \frac{u_1 - u_1 F_3(t)}{1 - F_3(t)} = u_1$$

that is,  $(t - T | X_1 \le t, X_2 \le t, X_3 > t)$  has the same distribution of  $(t - X_1 | X_1 \le t)$ . The representations for the other cases can be obtained in a similar way.

The property obtained in the preceding examples for  $T_t^W$  in the case of independent components and when  $W = \emptyset$  is actually a general property that can be stated as follows. **Proposition 3** If T is the lifetime of a coherent system with independent components, then  $\overline{Q}_t^{\emptyset} = Q_{\perp}$ .

The proof is obtained from (2.12) by replacing C with the product copula.

An immediate consequence of the previous proposition is described in the following statement. The proof is straightforward and therefore omitted.

**Corollary 1** If the components are independent, then  $(t - T|X_W > t, X^{W^c} \le t)$  has the same distribution of  $(t - T^*|X^{W^c} \le t)$ , where  $T^*$  is the lifetime of the system obtained from the original one by deleting the cut sets which have at least an element in W (i.e.,  $T^* = \min_{\{j:C_j \cap W = \emptyset\}} X^{C_j}$ ).

## **3** Stochastic comparisons

First, we briefly recall the definitions of the stochastic orders that will be used throughout this paper to compare random lifetimes or inactivity times. Let X and Y be two absolutely continuous random variables having a common support  $(0,\beta)$ , for a  $\beta \in \mathbb{R} \cup \{\infty\}$ , distribution functions F and G, reliability (survival) functions  $\overline{F} = 1 - F$  and  $\overline{G} = 1 - G$  and density functions f and g, respectively. Then we say that X is smaller than Y:

- in the stochastic order (denoted by  $X \leq_{ST} Y$ ) if  $\overline{F} \leq \overline{G}$  in  $(0, \beta)$ ;
- in the hazard rate order (denoted by  $X \leq_{HR} Y$ ) if the ratio  $\overline{G}/\overline{F}$  is increasing in  $(0,\beta)$ ;
- in the reversed hazard rate order (denoted by  $X \leq_{RHR} Y$ ) if the ratio G/F is increasing in  $(0, \beta)$ ;
- in the likelihood ratio order (denoted by  $X \leq_{LR} Y$ ) if the ratio g/f is increasing in  $(0, \beta)$ ;
- in the mean residual life order (denoted by  $X \leq_{MRL} Y$ ) if, and only if,  $E[X^t] \leq E[Y^t]$  for all  $t \in (0, \beta)$ .

We address the reader to [27] for a detailed description of these stochastic orders and to [1] for a list of examples of applications in the reliability theory. Here, in particular, we just point out that:

- $X \leq_{HR} Y$  if, and only if,  $(X t | X > t) \leq_{ST} (Y t | Y > t)$  for all  $t \in (0, \beta)$ ,
- $X \leq_{RHR} Y$  if, and only if,  $(t X | X \leq t) \geq_{ST} (t Y | Y \leq t)$  for all  $t \in (0, \beta)$
- $X \leq_{LR} Y$  if, and only if,  $(X \mid a \leq X \leq b) \leq_{ST} (Y \mid a \leq X \leq b)$  for all  $0 \leq a \leq b \leq \beta$ .

Hence, the hazard rate order and the reversed hazard rate order are equivalent to compare residual lifetimes and inactivity times of systems, respectively, at any age  $t \ge 0$ . Analogously, the likelihood ratio order can be used to compare both residual lifetimes and inactivity times, while this is not the case for the weaker stochastic order. Moreover, the following relationships are well known:

$$\begin{array}{cccc} X \leq_{LR} Y \implies X \leq_{HR} Y \implies & X \leq_{MRL} Y \\ & \downarrow & \downarrow & \downarrow \\ X \leq_{RHR} Y \implies X \leq_{ST} Y \implies E(X) \leq E(Y). \end{array}$$

In the previous section we have obtained representations for  $T_t$  and  $T_t^W$  as generalized distorted distributions based on the same baseline distributions. Now we can use these representations, and the results for generalized distorted distributions described in [16,18,19], to compare the inactivity times  $T_t$  and  $T_t^W$  for any W. We can also compare  $T_t^W$  for different sets W or inactivity times for different system structures. For it, we first recall some useful properties proved in recent literature.

In the case of distorted distributions (i.e., n = 1 in (2.3)), we have the following ordering properties, extracted from Theorems 2.4 and 2.5 in [16] and Theorem 2.3 in [19].

**Proposition 4** Let  $F_{q_1} = q_1(F)$  and  $F_{q_2} = q_2(F)$  be two distorted distributions (of two random variables  $X_1$  and  $X_2$ ) based on the same distribution function F and on the distortion functions  $q_1$  and  $q_2$ , respectively. Let  $\overline{q}_1$  and  $\overline{q}_2$  be the respective dual distortion functions. Then:

- (i)  $X_1 \leq_{ST} X_2$  for all F if and only if  $\overline{q}_1 \leq \overline{q}_2$  in (0,1).
- (ii)  $X_1 \leq_{HR} X_2$  for all F if and only if  $\overline{q}_2/\overline{q}_1$  is decreasing in (0,1).
- (iii)  $X_1 \leq_{RHR} X_2$  for all F if and only if  $q_2/q_1$  is increasing in (0,1).
- (iv)  $X_1 \leq_{LR} X_2$  for all F if and only if  $\overline{q}'_2/\overline{q}'_1$  is decreasing in (0,1).
- (v) If there exists  $u_0 \in (0,1]$  such that  $\overline{q}_2/\overline{q}_1$  is decreasing in  $(0,u_0)$  and increasing in  $(u_0,1)$ , then  $X_1 \leq_{MRL} X_2$  for all F such that the means of the respective distorted distributions are ordered (in the same sense).

In the general case (i.e. for generalized distorted distributions) we have the following results, extracted from Proposition 2.2 in [18].

**Proposition 5** Let  $F_{Q_1} = Q_1(F_1, \ldots, F_n)$  and  $F_{Q_2} = Q_2(F_1, \ldots, F_n)$  be two generalized distorted distributions (of two random variables  $X_1$  and  $X_2$ ) based on the same distribution functions  $F_1, \ldots, F_n$  and on the generalized distortion functions  $Q_1$  and  $Q_2$ , respectively. Let  $\overline{Q}_1$  and  $\overline{Q}_2$  be the respective generalized dual distortion functions. Then:

- (i)  $X_1 \leq_{ST} X_2$  for all  $F_1, \ldots, F_n$  if and only if  $\overline{Q}_1 \leq \overline{Q}_2$  in  $(0,1)^n$ .
- (ii)  $X_1 \leq_{HR} X_2$  for all  $F_1, \ldots, F_n$  if and only if  $\overline{Q}_2/\overline{Q}_1$  is decreasing in  $(0, 1)^n$ .
- (iii)  $X_1 \leq_{RHR} X_2$  for all  $F_1, \ldots, F_n$  if and only if  $Q_2/Q_1$  is increasing in  $(0,1)^n$ .

Note that both propositions provide necessary and sufficient conditions to obtain distributionfree orderings (except in the case of the mrl order). Now it is immediate to obtain the corresponding results to get distribution-free comparisons between  $T_t$  and  $T_t^W$ . Note that we can also compare  $T_t^W$ and  $T_t^{W^*}$  for different W and  $W^*$ . For example, the results to compare  $T_t$  and  $T_t^W$  can be stated as follows. The proofs are immediate from representations (2.9) and (2.10) and Propositions 4 and 5.

**Proposition 6** Let T be the lifetime of a coherent system with components having a common continuous distribution function F. Then:

- (i)  $T_t \leq_{ST} T_t^W$  ( $\geq_{ST}$ ) for all F if and only if  $\overline{q}_t \leq \overline{q}_t^W$  ( $\geq$ ) in [0,1].
- (ii)  $T_t \leq_{HR} T_t^W$  ( $\geq_{HR}$ ) for all F if and only if  $\overline{q}_t^W/\overline{q}_t$  is decreasing (increasing) in (0,1).
- (iii)  $T_t \leq_{RHR} T_t^W$  ( $\geq_{RHR}$ ) for all F if and only if  $q_t^W/q_t$  is increasing (decreasing) in (0,1).
- (iv)  $T_t \leq_{LR} T_t^W$  ( $\geq_{LR}$ ) for all F if and only if  $(q_t^W)'/q_t'$  is decreasing (increasing) in (0,1).
- (v) If there exists  $u_0 \in (0,1]$  such that  $\overline{q}_t^W/\overline{q}_t$  is decreasing (increasing) in  $(0,u_0)$  and increasing (decreasing) in  $(u_0,1)$ , then  $T_t \leq_{MRL} T_t^W$  ( $\geq_{MRL}$ ) for all F such that  $E(T_t) \leq E(T_t^W)$  ( $\geq$ ).

**Proposition 7** Let T be the lifetime of a coherent system with components having distribution functions  $F_1, \ldots, F_n$ . Then:

- (i)  $T_t \leq_{ST} T_t^W$  ( $\geq_{ST}$ ) for all  $F_1, \ldots, F_n$  if and only if  $\overline{Q}_t \leq \overline{Q}_t^W$  ( $\geq$ ) in  $(0,1)^n$ . (ii)  $T_t \leq_{HR} T_t^W$  ( $\geq_{HR}$ ) for all  $F_1, \ldots, F_n$  if and only if  $\overline{Q}_t^W / \overline{Q}_t$  is decreasing (increasing) in  $(0,1)^n$ .
- (iii)  $T_t \leq_{RHR} T_t^W$  ( $\geq_{RHR}$ ) for all  $F_1, \ldots, F_n$  if and only if  $Q_t^W/Q_t$  is increasing (decreasing) in  $(0,1)^n$ .

A simple example of application of the previous results, dealing with the comparison of inactivity times  $T_t = (t - T | T \le t)$  for series and parallel systems of two components, is given now.

Example 3 Consider two components having possibly dependent lifetimes  $X_1$  and  $X_2$ , with the same distribution F, and consider  $T^{\max} = \max(X_1, X_2)$  and  $T^{\min} = \min(X_1, X_2)$ , lifetimes of the corresponding parallel and series system. It is rather intuitive, and actually easy to analytically verify, that if the components have independent lifetimes then the inactivity times  $T_t^{\max}$  and  $T_t^{\min}$ are comparable in the likelihood order, i.e., it holds  $T_t^{\max} \leq_{LR} T_t^{\min}$  for any t > 0. However, using Proposition 4,(iv) one can verify that this inequality does not necessary hold for any dependence structure (copula) of the vector  $(X_1, X_2)$ .

In fact, denoting with C the copula of the vector  $(X_1, X_2)$ , one has

$$P(t - T^{\min} > x | T^{\min} \le t) = \overline{q}_t^{\min}(\overline{F}_t(x)),$$

where  $\overline{F}_t(x) = F(t-x)/F(t)$  and

$$\overline{q}_t^{\min}(u) = \frac{2uF(t) - C(uF(t), uF(t))}{2F(t) - C(F(t), F(t))} = \frac{2uF(t) - \delta(uF(t))}{2F(t) - \delta(F(t))}$$

being  $\delta$  the diagonal section of the copula C. Similarly,

$$P(t - T^{\max} > x | T^{\max} \le t) = \overline{q}_t^{\max}(\overline{F}_t(x)),$$

where

$$\overline{q}_t^{\max}(u) = \frac{C(uF(t), uF(t))}{C(F(t), F(t))} = \frac{\delta(uF(t))}{\delta(F(t))}.$$

Observe that, by Proposition 4, (iv), the inequality  $T_t^{\max} \leq_{LR} T_t^{\min}$  holds if and only if

$$\frac{d \ \overline{q}_t^{\min}(u)/du}{d \ \overline{q}_t^{\max}(u)/du} = \frac{\delta(F(t))}{2F(t) - \delta(F(t))} \ \frac{2F(t) - F(t)\delta'(uF(t))}{F(t)\delta'(uF(t))}$$

is decreasing in u, thus if

$$\frac{2-\delta'(uF(t))}{\delta'(uF(t))}$$

is decreasing in u. The latter is satisfied if and only if  $\delta(u)$  is convex in (0,1). A list of copulas have convex diagonal section (such as: Marshall-Olkin for any value of the parameters, FGM with negative value of the parameter  $\theta$ , i.e.,  $\theta \in (-1,0]$ , Gumbel copulas, Clayton and other Archimedean copula, etc.). However, do not exist copulas having a concave diagonal section. A copula whose diagonal section is neither convex nor concave is the FGM with  $\theta \in (0, 1]$ . Thus, for this copula the stated property does not hold. Note that for the product copula we have  $\delta(u) = u^2$  which is a convex function. So the stated property holds in the case of independent components. Proceeding in a similar way and by using Proposition 4, (ii), one can prove that  $T_t^{\max} \leq_{HR} T_t^{\min}$  holds for any t > 0 if and only if  $\delta(u)/u$  is increasing in [0, 1].  Under the assumption of independence between components' lifetimes, a simple proof of the inequality (1.1) mentioned in the Introduction follows by a direct application of Proposition 7. Using this result, in fact, it is possible to prove the stochastic comparisons between the inactivity time of a system conditioning on the fact that it failed before a time t or that all its components have failed before time t.

**Proposition 8** If T is the lifetime of a coherent system formed by n components having independent lifetimes  $X_1, \ldots, X_n$ , then

$$(t - T|T < t) \leq_{ST} (t - T|X_1 < t, \dots, X_n < t) \quad \forall t \ge 0.$$
(3.1)

Proof Let  $F_1, \ldots, F_n$  denote the distribution functions of  $X_1, \ldots, X_n$ . From (2.9), the reliability function of  $T_t = (t - T | T \le t)$  can be written as

$$\overline{F}_t(x) = \overline{Q}_t(\overline{F}_{1,t}(x), \dots, \overline{F}_{n,t}(x))$$

for  $x \in [0,t]$ , where  $\overline{F}_{i,t}(x) = F_i(t-x)/F_i(t)$  is the reliability function of the inactivity time  $X_{i,t} = (t - X_i | X_i \le t)$  of the *i*th component for i = 1, ..., n, and where

$$\overline{Q}_t(u_1,\ldots,u_n) = \frac{Q(u_1F_1(t),\ldots,u_nF_n(t))}{Q(F_1(t),\ldots,F_n(t))}$$

is a generalized distortion function.

On the other hand, from Proposition 3, the reliability function  $T_t^{\emptyset} = (t - T | X_1 < t, \dots, X_n < t)$  can be written as

$$\overline{F}_t(x) = Q(\overline{F}_{1,t}(x), \dots, \overline{F}_{n,t}(x))$$

for  $x \in [0, t]$ , where Q is the generalized distortion function of T in the case of independent components.

Therefore, from Proposition 7(i),  $T_t \leq_{ST} T_t^{\emptyset}$  holds for all  $F_1, \ldots, F_n$ , if and only if

$$\frac{Q(u_1F_1(t),\ldots,u_nF_n(t))}{Q(F_1(t),\ldots,F_n(t))} \le Q(u_1,\ldots,u_n)$$

holds for all  $u_1, \ldots, u_n \in [0, 1]$ . Hence,  $T_t \leq_{ST} T_t^{\emptyset}$  holds for all  $F_1, \ldots, F_n$ , and for all t, if and only if

$$Q(u_1v_1, \dots, u_nv_1) \le Q(u_1, \dots, u_n)Q(v_1, \dots, v_n)$$
(3.2)

for all  $u_1, \ldots, u_n, v_1, \ldots, v_n \in [0, 1]$ . Now we use the fact that Q is the reliability structure (dual generalized distortion) function of the dual system (since the minimal path sets of the dual systems are the minimal cut sets of T and  $C = \overline{C}$  in the case of independent components). Moreover, it is well known that (3.2) holds for reliability structure functions in the case of independent components (see (5.2) in [1], p. 183). This completes the proof.

Actually, even if the stochastic inequality (3.1) is in general false, Proposition 8 can be generalized to systems with dependent components under additional assumptions on the structure of dependence among them.

**Proposition 9** If T is the lifetime of a coherent system formed by n components having lifetimes  $(X_1, \ldots, X_n)$  with a continuous joint distribution such that

$$(t - X_A | X_A < t, X^{A^c} \ge t) \le_{ST} (t - X_A | X_1 < t, \dots, X_n < t)$$
(3.3)

for all nonempty set  $A \subseteq \{1, \ldots, n\}$ , then

$$(t - T|T < t) \leq_{ST} (t - T|X_1 < t, \dots, X_n < t)$$
 for all  $t > 0$ .

*Proof* Denote with  $\mathfrak{C}_1, \ldots, \mathfrak{C}_s$ , all the cut sets of the system where  $\mathfrak{C}_s = \{1, \ldots, n\}$  (note that here we consider all the cut sets, and not only the minimal ones). Then it holds that

$$\{T < t\} = \bigcup_{i=1}^{s} \{X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t\}$$

for any  $x \ge 0$ . Note that it is a union of disjoint events.

For any  $x, t \ge 0$ , let

$$a_i = \Pr(X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t) \text{ and } b_i = \Pr(T < t - x, X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t).$$

Then we have

$$\begin{aligned} \Pr(T_t > x) &= \Pr(t - T > x | T < t) = \frac{\Pr(T < t - x, T < t)}{\Pr(T < t)} \\ &= \frac{\sum_{i=1}^s \Pr(T < t - x, X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t)}{\sum_{i=1}^s \Pr(X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t)} = \frac{\sum_{i=1}^s b_i}{\sum_{i=1}^s a_i}.\end{aligned}$$

The lifetime T can be written as  $T = \tau(X_1, \ldots, X_n)$ . For any cut set  $\mathfrak{C}_i$ , let  $T_i = \tau_i(X_{\mathfrak{C}_i})$  be the lifetime of the system obtained from T by assuming that all the components not included in  $\mathfrak{C}_i$  are always working. Of course, we have  $T \leq T_i$  for all i.

Then for any  $i = 1, \ldots, s$ 

$$\begin{aligned} \frac{b_i}{a_i} &= \Pr(T < t - x | X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t) = \Pr(\tau(X) < t - x | X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t) \\ &= \Pr(\tau_i(X_{\mathfrak{C}_i}) < t - x | X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t) = \Pr(\tau_i(t - X_{\mathfrak{C}_i}) > x | X_{\mathfrak{C}_i} < t, X^{\mathfrak{C}_i^c} \ge t) \\ &\leq \Pr(\tau_i(t - X_{\mathfrak{C}_i}) > x | X_{\mathfrak{C}_s} < t) = \Pr(t - T_i > x | X_{\mathfrak{C}_s} < t) \\ &\leq \Pr(t - T > x | X_{\mathfrak{C}_s} < t) = \Pr(T < t - x | X_{\mathfrak{C}_s} < t) = \frac{b_s}{a_s}, \end{aligned}$$

where the first inequality is obtained from assumption (3.3) and the second from  $T \leq T_i$ . Thus  $a_s b_i \leq a_i b_s$  and

$$a_s b_1 + \dots + a_s b_s \le a_1 b_s + \dots + a_s b_s.$$

Hence

$$\Pr(t - T > x | T < t) = \frac{\sum_{i=1}^{s} b_i}{\sum_{i=1}^{s} a_i} \le \frac{b_s}{a_s} = \Pr(t - T > x | X_{\mathfrak{C}_s} < t)$$

and the stated result holds.

An example where the assumptions of the previous proposition are satisfied for any non empty set  $A \subseteq I$  and  $t \ge 0$  is when the vector of lifetimes  $(X_1, \ldots, X_n)$  has an MTP2 joint density f, i.e., when f satisfies  $f(\boldsymbol{x})f(\boldsymbol{y}) \le f(\boldsymbol{x} \lor \boldsymbol{y})f(\boldsymbol{x} \land \boldsymbol{y})$  for any  $\boldsymbol{x}, \boldsymbol{y} \in \mathbb{R}^n$ . See, e.g., [6] or [9] for the formal definition and examples of random vectors satisfying the MTP2 property. The proof of this assertion is similar to the proof of Theorem 11.2.2 in [12], and therefore omitted.

Further simple conditions may be stated for the case of series systems with two dependent components, as described in the following statement.

**Proposition 10** If  $T = \min(X_1, X_2)$  and the copula C of  $(X_1, X_2)$  satisfies the conditions

$$zC(x,y) \ge yC(x,z), \text{ for all } 0 \le x \le 1, \ 0 \le y \le z \le 1,$$
(3.4)

and

$$zC(x,y) \ge xC(z,y), \text{ for all } 0 \le x \le z \le 1, \ 0 \le y \le 1,$$
 (3.5)

then  $(t - T | T \le t) \le_{ST} (t - T | X_1 \le t, X_2 \le t).$ 

Proof From Example 1, the dual distortion functions of  $T_t = (t - T | T \le t)$  and  $T_t^{\emptyset} = (t - T | X_1 \le t, X_2 \le t)$  are, respectively,

$$\overline{Q}_t(u_1, u_2) = \frac{u_1 F_1(t) + u_2 F_2(t) - C(u_1 F_1(t), u_2 F_2(t))}{F_1(t) + F_2(t) - C(F_1(t) F_2(t))}$$

and

$$\overline{Q}_t^{\emptyset}(u_1, u_2) = \frac{C(u_1F_1(t), F_2(t)) + C(F_1(t), u_2F_2(t)) - C(u_1F_1(t), u_2F_2(t))}{C(F_1(t), F_2(t))}$$

Thus, the stated result holds if and only if these two distortion functions satisfy

$$\overline{Q}_t(u_1, u_2) \le \overline{Q}_t^{\emptyset}(u_1, u_2) \tag{3.6}$$

By taking  $x = F_1(t)$ ,  $y = u_2 F_2(t)$  and  $z = F_2(t)$  in (3.4) we get

$$F_2(t)C(F_1(t), u_2F_2(t)) \ge u_2F_2(t)C(F_1(t), F_2(t)).$$

Analogously, by taking  $x = u_1 F_1(t)$ ,  $y = F_2(t)$  and  $z = F_1(t)$  in (3.5) we get

$$F_1(t)C(u_1F_1(t), F_2(t)) \ge u_1F_1(t)C(F_1(t), F_2(t)).$$

Hence (3.6) holds if

$$C(F_1(t), F_2(t))C(u_1F_1(t), u_2F_2(t)) + F_1(t)C(F_1(t), u_2F_2(t)) - F_1(t)C(u_1F_1(t), u_2F_2(t)) - C(F_1(t), F_2(t))C(F_1(t), u_2F_2(t)) \ge 0$$

and

$$C(F_1(t), F_2(t))C(u_1F_1(t), u_2F_2(t)) + F_2(t)C(u_1F_1(t), F_2(t)) - F_2(t)C(u_1F_1(t), u_2F_2(t)) - C(F_1(t), F_2(t))C(u_1F_1(t), F_2(t)) \ge 0$$

hold. The first term can be written as

$$(F_1(t) - C(F_1(t), F_2(t)))(C(F_1(t), u_2F_2(t)) - C(u_1F_1(t), u_2F_2(t)))$$

and the second one as

$$(F_2(t) - C(F_1(t), F_2(t)))(C(u_1F_1(t), F_2(t)) - C(u_1F_1(t), u_2F_2(t))).$$

Hence both terms are nonnegative since C is increasing and C(u, 1) = C(1, u) = u.

Remark 1 Conditions (3.4) and (3.5) can be seen as positive dependence properties (weaker than  $TP_2$  property). In fact, letting  $x = F_1(t), z = F_2(t)$  and  $y = F_2(s)$ , with  $s \leq t$ , one can immediately observe that (3.4) is equivalent to

$$\Pr(X_1 < t | X_2 < s) \ge \Pr(X_1 < t | X_2 < t), \text{ for all } s \le t,$$
(3.7)

and, similarly, one can see that (3.5) is equivalent to

$$\Pr(X_2 < t | X_1 < s) \ge \Pr(X_2 < t | X_1 < t), \text{ for all } s \le t.$$
(3.8)

Hence, (3.4) and (3.5) hold if  $\Pr(X_1 < t | X_2 < s)$  and  $\Pr(X_2 < t | X_1 < s)$  are decreasing in s for  $s \leq t$ , i.e., if  $X_2$  is Left Tail Decreasing (LTD) in  $X_1$  and  $X_1$  is LTD in  $X_2$ . The LTD notion is a well know property describing positive dependence among random variables; see, e.g., [24], Chapter 5, or [3] on its formal definition and applications in modeling dependence.

## 4 Illustrative examples

A list of examples of applications of the theoretical results described in previous sections are provided here. The first one proves that the ordering in (1.1) is not always true.

*Example 4* Consider the lifetime  $T = \min(X_1, X_2)$  of a series system formed by two components having non independent lifetimes  $X_1$  and  $X_2$ . Observe that for this system we have

$$\Pr(t - T \le x | T \le t) = \frac{\Pr(t - x \le \min(X_1, X_2) \le t)}{\Pr(\min(X_1, X_2) \le t)} = \frac{\overline{F}(t - x, t - x) - \overline{F}(t, t)}{1 - \overline{F}(t, t)} = p_{1,t}(x)$$

and

$$\Pr(t - T \le x | X_1, X_2 \le t) = \frac{\overline{F}(t - x, t - x) - \overline{F}(t - x, t) - \overline{F}(t, t - x) + \overline{F}(t, t)}{1 - \overline{F}(0, t) - \overline{F}(t, 0) + \overline{F}(t, t)} = p_{2,t}(x),$$

where  $\overline{F}$  denotes the joint reliability function of  $(X_1, X_2)$ .

Assume now that the pair  $(X_1, X_2)$  has a Gumbel's bivariate exponential distribution, i.e., let

$$\overline{F}(x_1, x_2) = \Pr(X_1 > x_1, X_2 > x_2) = \exp(-\alpha_1 x_1 - \alpha_2 x_2 - \theta \alpha_1 \alpha_2 x_1 x_2), \quad \alpha_i \ge 0, \quad \theta \in (0, 1)$$

It can be numerically verified that in this case the inequality  $p_{1,t}(x) \leq p_{2,t}(x)$  does not holds for all  $x \leq t$  (see, e.g., Figure 1, in which  $\alpha_1 = 4$ ,  $\alpha_2 = 1$ ,  $\theta = 1/2$  and t = 1), i.e., inequality (1.1) does not hold for all t > 0.

Two examples where Proposition 10 can be applied are described now.

Example 5 Recall that a copula C is called Archimedean if it can be written in the form

$$C(u_1, u_2) = \phi^{-1}(\phi(u_1) + \phi(u_2)), \tag{4.1}$$

for any continuous and strictly decreasing function  $\phi : [0, 1] \mapsto [0, \infty]$  such that  $\phi(1) = 0$ . In this case, the function  $\phi$  is called generator of the Archimedean copula. In [2], Proposition 6.1, it is proved that any Archimedean copula is Totally Positive of order 2  $(TP_2)$ , i.e., it satisfies

$$C(x_1, y_1)C(x_2, y_2) \ge C(x_1, y_2)C(x_2, y_1) \quad \forall \ 0 \le x_1 \le x_2 \le 1 \text{ and } 0 \le y_1 \le y_2 \le 1,$$
(4.2)

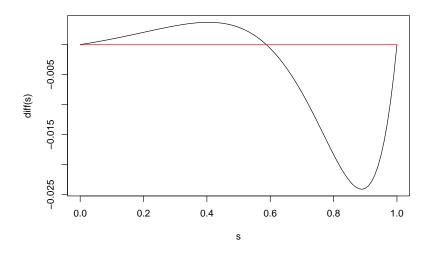


Fig. 1 Plot of the difference  $Pr(t - T \le x | X_1, X_2 \le t) - Pr(t - T \le x | T \le t)$  for  $T = min(X_1, X_2)$  when t = 1 and the vector  $(X_1, X_2)$  has Gumbel's bivariate exponential distribution with  $\alpha_1 = 4$ ,  $\alpha_2 = 1$  and  $\theta = 1/2$ .

if the inverse  $\phi^{-1}$  of its generator function is log-convex. Examples of Archimedean copulas for which (4.2) holds are, for example, the Clayton or the Gumbel-Hougaard copulas, for any values of their parameters. Ali-Mikhail-Haq copulas, for positive values of the parameter, also satisfy this property.

Let now  $T = \min(X_1, X_2)$  be the lifetime of a series system with two dependent components having an Archimedean copula C. Since the property (4.2) clearly implies (3.4) and (3.5), then (3.6) holds whenever C is Archimedean with log-convex inverse generator  $\phi^{-1}$ .

Example 6 Let  $T = \min(X_1, X_2)$  be the lifetime of a series system with two dependent components having dependent lifetimes  $X_1$  and  $X_2$  connected by a Farlie-Gumbel-Morgenstern (FGM) copula, i.e., the copula defined as  $C(x, y) = xy[1 + \alpha(1 - x)(1 - y)]$  with  $-1 \le \alpha \le 1$ . For  $0 \le \alpha < 1$  it is easy to verify that

$$zxy[1 + \alpha - \alpha x - \alpha y + \alpha xy] \ge xyz[1 + \alpha - \alpha x - \alpha z + \alpha xz], \quad \text{for } y \le z$$

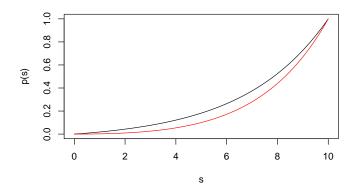
and

$$zxy[1 + \alpha - \alpha x - \alpha y + \alpha xy] \ge xzy[1 + \alpha - \alpha y - \alpha z + \alpha yz], \quad \text{for } x \le z.$$

Thus, both (3.4) and (3.5) hold, and so (3.6) holds too.

The next example shows that (3.7) and (3.8) are not necessary conditions for the stochastic comparison (3.1) where  $T = \min(X_1, X_2)$ .

Example 7 Let  $(X_1, X_2)$  have Gumbel's bivariate exponential distribution as seen in Example 4, with  $\alpha_1 = \alpha_2 = 1$ ,  $\theta = 0.5$ . The copula *C* of this vector does not satisfy conditions (3.4) and (3.5), otherwise inequality  $(t - T|T < t) \leq_{ST} (t - T|X_1 < t, X_2 < t)$  would be satisfied in Example 4, by Proposition 10. However, for this particular choice of the parameters  $\alpha_i$  and  $\theta$ , and letting t = 1, the stochastic inequality  $(t - T|T < t) \leq_{ST} (t - T|X_1 < t, X_2 < t)$  is satisfied. This can be verified numerically: Figure 2 shows the plots of  $\Pr(t - T \leq x | T \leq t)$  (black) and  $\Pr(t - T \leq x | X_1, X_2 \leq t)$ (red) for  $x \in (0, 1]$ .



**Fig. 2** Plots of  $Pr(t - T \le x | T \le t)$  (black) and  $Pr(t - T \le x | X_1, X_2 \le t)$  (red), for  $\alpha_1 = \alpha_2 = 1$ ,  $\theta = 0.5$  and t = 1, with  $x \in (0, 1]$ , when  $(X_1, X_2)$  is described as in Example 7.

Example 8 Consider a k-out-of-n system whose lifetime T corresponds to the k-th failure of a component, and let the  $X_i$ , i = 1, ..., n be the lifetimes of the n components. Assuming that the  $X_i$ are independent ad identically distributed, with cumulative distribution F, by Proposition 9 follows that the inactivity time  $T_t$  under the general condition that the system is failed in t is always stochastically bounded by  $(t - T|X_1 < t, ..., X_n < t)$  that is,  $\forall x \in (0, t)$  it holds

$$Pr(t - T > x|T < t) \le Pr(t - T > x|X_{n:n} < t)$$
  
= 
$$\frac{Pr(X_{k:n} < t - x, X_{n:n} < t)}{Pr(X_{n:n} < t)} = \frac{\sum_{j=k}^{n} {n \choose j} F^{j}(t - x) F^{n-j}(t)}{F^{n}(t)}$$

We conclude the section observing that a statement somehow related to Proposition 8 has been provided in [11], where, in Theorem 4, it is proved that

$$(\max\{X_1, X_2\})_t \leq_{HR} \max\{X_{1,t}, X_{2,t}\} \text{ and } (\min\{X_1, X_2\})_t \geq_{HR} \min\{X_{1,t}, X_{2,t}\}$$
(4.3)

for independent components having lifetimes  $X_1$  and  $X_2$ , and inactivity times  $X_{1,t}$  and  $X_{2,t}$ . Actually, Proposition 8 is clearly different, since the inequalities in (4.3) compare the inactivity time of systems with the maximum, or minimum, among inactivity times of their components. Moreover, the following example proves that inequality (3.1) does not holds in general for the  $\leq_{HR}$  order. However, it also shows that (3.1) can be satisfied for the  $\leq_{LR}$  order whenever T is the lifetime of a series system having independent and identically distributed lifetimes of the components.

*Example 9* Let us consider a series system with two independent components and lifetime  $T = \min(X_1, X_2)$ . From Example 1, the dual distortion functions of  $T_t$  and  $T_t^{\emptyset}$  are

$$\overline{Q}_t(u_1, u_2) = \frac{u_1 F_1(t) + u_2 F_2(t) - u_1 u_2 F_1(t) F_2(t)}{F_1(t) + F_2(t) - F_1(t) F_2(t)}$$

and

$$\overline{Q}_t^{\emptyset}(u_1, u_2) = Q(u_1, u_2) = u_1 + u_2 - u_1 u_2$$

From Proposition 8, we know that  $T_t \leq_{ST} T_t^{\emptyset}$  holds for all t and all continuous distributions  $F_1, F_2$ . To study if  $T_t \leq_{HR} T_t^{\emptyset}$  holds, we consider the ratio

$$R_t(u_1, u_2) = \frac{\overline{Q}_t^{\emptyset}(u_1, u_2)}{\overline{Q}_t(u_1, u_2)} = \frac{(u_1 + u_2 - u_1 u_2)(F_1(t) + F_2(t) - F_1(t)F_2(t))}{u_1 F_1(t) + u_2 F_2(t) - u_1 u_2 F_1(t)F_2(t)}.$$

It can be seen that if  $F_1(t) = 0.5$ ,  $F_2(t) = 0.7$ , then this ratio is increasing in  $u_1$  when  $u_2 = 0.1$  and it is decreasing when  $u_2 = 0.9$ . Therefore  $T_t \leq_{HR} T_t^{\emptyset}$  does not hold.

However, if we assume that the components are IID (i.e.,  $F_1 = F_2 = F$ ), then we should study the ratio

$$r_t(u) = \frac{\overline{q}_t^{\emptyset}(u)}{\overline{q}_t(u)} = \frac{(2u - u^2)(2F(t) - F^2(t))}{2uF(t) - u^2F^2(t)} = \frac{(2 - u)(2 - F(t))}{2 - uF(t)}$$

whose derivative satisfies

$$r_t'(u) =_{sign} -2 + 2F(t) \le 0.$$

Therefore,  $r_t$  is decreasing and  $T_t \leq_{HR} T_t^{\emptyset}$  holds for all t and for all IID components. Even more, to study if  $T_t \leq_{LR} T_t^{\emptyset}$  holds we consider the ratio

$$g_t(u) = \frac{(q_t^{\emptyset})'(u)}{q_t'(u)} = \frac{(1-u)(2-F(t))}{1-uF(t)}$$

whose derivative satisfies

$$q_t'(u) =_{sign} -1 + F(t) \le 0.$$

Therefore,  $g_t$  is decreasing and  $T_t \leq_{LR} T_t^{\emptyset}$  holds for all t and for all IID components.

## Acknowledgements

We sincerely thanks the two referees for their careful reading of the manuscript and for their thoughtful comments and suggestions that greatly improved the presentation of the paper.

JN is partially supported by Ministerio de Economía y Competitividad under grant MTM2012-34023-FEDER. FP and ML are partially supported by the Italian INDAM - Istituto Nazionale Di Alta Matematica.

## References

- 1. Barlow, R.E. & Proschan, F. (1975). Statistical Theory of Reliability and Life Testing. Probability Models. International Series in Decision Processes. Holt, Rinehart and Winston, Inc., New York.
- Bassan, B. & Spizzichino, F. (2005). Relations among univariate aging, bivariate aging and dependence for exchangeable lifetimes. *Journal of Multivariate Analysis* 93:313–339.
- Colangelo, A., Scarsini, M. & Shaked, M. (2005). Some notions of multivariate positive dependence. *Insurance: Mathematics and Economics* 37:13–26.
- Durante, F., Saminger-Platz, S. & Sarkoci, P. (2008). On representations of 2-increasing binary aggregation functions. *Information Sciences* 178:4534–4541.
- 5. Durante, F. & Sempi, C. (2015). Principles of copula theory. CRC/Chapman & Hall, London.
- Fang. Z & Hu, T. (1997). Developments on MTP<sub>2</sub> properties of absolute value multinormal variables with nonzero means. Acta Mathematicae Applicatae Sinica 13:376–384.
- 7. Goli, S & Asadi, M. (2017). A study on the conditional inactivity time of coherent systems. Metrica 80:227–241.
- 8. Grabisch, M., Marichal, J.-L. Mesiar, R. & Pap, E. (2009). Aggregation functions. Cambridge University Press.
- Karlin, S. & Rinott, Y. (1980). Classes of orderings of measures and related correlation inequalities. I. Multivariate totally positive distributions. *Journal of Multivariate Analysis*. 10:467–498.

- 10. Kuo, W. & Zhu, X. (2012). Importance Measures in Reliability, Risk, and Optimization: Principles and Applications. Wiley.
- Li, X. & Lu, J. (2003). Stochastic comparisons on residual life and inactivity time of series and parallel systems. Probability in the Engineering and Informational Sciences 17:267–275.
- Li, X., Pellerey, F. & You, Y. (2013). On used systems and systems with used components. In: Stochastic orders in Reliability and risk. In honor of Professor Moshe Shaked. Springer, New York, pp. 219–233.
- Navarro, J. (2016). Distribution-free comparisons of residual lifetimes of coherent systems based on copula properties. To appear in Statistical Papers. Published online first June 2016 with DOI 10.1007/s00362-016-0789-0.
- 14. Navarro, J. (2016). Stochastic comparisons of generalized mixtures and coherent systems. Test 25: 150–169.
- 15. Navarro J. & Durante F. (2015). Copula-based representations for the reliability of the residual lifetimes of coherent systems with dependent components. Submitted.
- Navarro, J., del Aguila, Y., Sordo, M.A. & Suarez-Llorens, A. (2013). Stochastic ordering properties for systems with dependent identically distributed components. *Applied Stochastic Models in Business and Industry* 29:264– 278.
- Navarro, J., del Aguila, Y., Sordo, M.A. & Suarez-Llorens, A. (2014). Preservation of reliability classes under the formation of coherent systems. *Applied Stochastic Models in Business and Industry* 30, 444–454.
- Navarro, J., del Aguila, Y., Sordo, M.A. & Suarez-Llorens, A. (2016). Preservation of stochastic orders under the formation of generalized distorted distributions. Applications to coherent systems. *Methodology and Computing* in Applied Probability 18, 2:529–545.
- Navarro, J., & Gomis, M.C. (2016). Comparisons in the mean residual life order of coherent systems with identically distributed components. *Applied Stochastic Models in Business and Industry* 32, 1:33–47.
- Navarro, J., Pellerey, F. & Di Crescenzo, A. (2015). Orderings of coherent systems with randomized dependent components. *European Journal of Operational Research* 240:127–139.
- Navarro, J. & Rychlik, T. (2010). Comparisons and bounds for expected lifetimes of reliability systems. *European Journal of Operational Research* 207:309–317.
- Navarro, J., Samaniego, F.J. & Balakrishnan, N. (2011). Signature-based representations for the reliability of systems with heterogeneous components. *Journal of Applied Probability* 48:856–867.
- Navarro, J. & Spizzichino, F. (2010). Comparisons of series and parallel systems with components sharing the same copula. Applied Stochastic Models in Business and Industry 26:775–791.
- 24. Nelsen, R.B. (2006). An introduction to copulas. Second edition. Springer Series in Statistics. Springer, New York.
- Pellerey, F. & Petakos, K. (2002). On the closure of the NBUC class under the formation of parallel systems. IEEE Transactions on Reliability 51:452–454.
- Samaniego, F.J., & Navarro, J. (2016). On comparing coherent systems with heterogeneous components. Advances in Applied Probability 48: 88–111.
- 27. Shaked, M. & Shanthikumar, J.G. (2007). Stochastic orders. Springer Series in Statistics. Springer, New York.
- Zhang Z. (2010). Mixture representations of inactivity times of conditional coherent systems and their applications. Journal of Applied Probability 47:876–885.
- 29. Zhang Z & Balakrishnan N (2016). Representations of the inactivity time for coherent systems with heterogeneous components and some ordered properties. *Metrika* 79:113–126.