# UNIVERSITY OF OSLO

# Faculty of Mathematics and Natural Sciences



Please make sure that your copy of the problem set is complete before you attempt to answer anything.

#### Problem 1

(a) Give the definitions of a multistate strongly coherent system, a multistate coherent system and a multistate weakly coherent system. Explain the mutual relations between these three.

(b) Give the definition of a binary type multistate monotone system. What is the idea behind this measure?

### Problem 2

(a) Assume that  $X_1, \ldots, X_n$  are associated random variables. Show that for  $j \in \mathbb{C}$  $\{1, \ldots, M\}$  we have for the multistate series system.

$$
P[\min_{1 \le i \le n} X_i \ge j] \ge \prod_{i=1}^n p_i^j \tag{1}
$$

(b) Let  $\phi$  be a multistate structure function that is nondecreasing in each argument and we assume that

$$
\min_{1 \le i \le n} x_i \le \phi(\boldsymbol{x}) \tag{2}
$$

Assume that  $X_1, \ldots, X_n$  are associated random variables. Show that we have

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$$
\prod_{i=1}^{n} p_i^j \le p_\phi^j \tag{3}
$$

Comment on the result.

#### Problem 3

Let X be a vector of environmental variables with sample space  $\mathcal{X} \subseteq \mathbb{R}^n$ . Furthermore let  $P_e \in (0, 0.5)$  be a given exceedence probability. We shall in this exercise look into how one can identify a convex set  $\mathcal{B} \subset \mathcal{X}$  such that for all hyperplanes  $\Pi$  that are tangents to B, we have  $P[X \in \Pi^+] = P_e$ , where  $\Pi^+$  denotes the halfspace that is bounded by the hyperplane  $\Pi$  and not containing  $\mathcal{B}$ . We also introduce  $\Pi^-$  denoting the halfspace that is complementary to  $\Pi^+$ , such that  $\mathcal{B} \subseteq \Pi^-$ . The boundary of  $\mathcal{B}$  is denoted by  $\partial \mathcal{B}$  and is what is called an environmental contour.

(a) Explain how such an environmental contour can be used to check whether a certain mechanical structure is safe.

We assume in the rest of the exercise that  $\mathbf{X} = (T, H)$ . For  $\theta \in [0, 2\pi)$  we define the random variable  $Y(\theta) = T \cos(\theta) + H \sin(\theta)$ . We also introduce

the function  $C(\theta)$  defined for  $\theta \in [0, 2\pi)$ :

$$
C(\theta) = \inf \{ C : P[T\cos(\theta) + H\sin(\theta) > C] = P_e \}.
$$

(b) What property does  $C(\theta)$  have related to the probability distribution of  $Y(\theta)$ ?

We also introduce:

$$
\Pi(\theta) = \{(t, h) : t \cos(\theta) + h \sin(\theta) = C(\theta)\},
$$
  
\n
$$
\Pi^+(\theta) = \{(t, h) : t \cos(\theta) + h \sin(\theta) > C(\theta)\},
$$
  
\n
$$
\Pi^-(\theta) = \{(t, h) : t \cos(\theta) + h \sin(\theta) \le C(\theta)\}.
$$

Then it can be shown that the set  $\beta$  can can be written as:

$$
\mathcal{B} = \bigcap_{\theta \in [0, 2\pi)} \Pi^-(\theta).
$$

We consider the intersection point  $(t(\theta_1, \theta_2), h(\theta_1, \theta_2))$  between the hyperplanes  $\Pi(\theta_1)$  $\Pi(\theta_2)$ .

(c) Show the first of the two following relations:

$$
t(\theta_1, \theta_2) = \frac{\sin(\theta_2)C(\theta_1) - \sin(\theta_1)C(\theta_2)}{\sin(\theta_2 - \theta_1)}
$$

$$
h(\theta_1, \theta_2) = \frac{-\cos(\theta_2)C(\theta_1) + \cos(\theta_1)C(\theta_2)}{\sin(\theta_2 - \theta_1)}
$$

Then let  $(t(\theta), h(\theta)) = \lim_{\delta \to 0^+} (t(\theta, \theta + \delta), h(\theta, \theta + \delta)).$ 

(d) Show that:

$$
\left(\begin{array}{c} t(\theta) \\ h(\theta) \end{array}\right) = \left[\begin{array}{cc} C(\theta) & -C'(\theta) \\ C'(\theta) & C(\theta) \end{array}\right] \cdot \left(\begin{array}{c} \cos{(\theta)} \\ \sin{(\theta)} \end{array}\right),
$$

where  $C'(\theta)$  denotes the derivative of  $C(\theta)$ , and explain in a short way how this expression can be used to identify  $\partial \mathcal{B}$ .

(e) What sort of contour is  $\partial \mathcal{B}$  if  $C(\theta) = R$  for all  $\theta \in [0, 2\pi)$ ? Give a reason for your answer.

Assume we have carried through a Monte Carlo simulation based on the joint probability distribution of  $(T, H)$ , and that we have generated n vectors:

$$
(T_1, H_1), \dots, (T_n, H_n) \tag{4}
$$

For a given  $\theta \in [0, 2\pi)$  we then calculate:

$$
Y_i(\theta) = T_i \cos(\theta) + H_i \sin(\theta), \quad i = 1, \dots, n
$$
\n<sup>(5)</sup>

(f) Explain how we can use  $Y_1(\theta), \ldots, Y_n(\theta)$  to estimate  $C(\theta)$ .

## Problem 4

Consider events occuring on a time axis starting from time  $t = 0$ . Let  $0 < S_1 < S_2 < \ldots$ be the event times and let  $T_i = S_i - S_{i-1}$  be the times between events, where  $S_0 = 0$ . Let further  $N(t)$  for  $t > 0$  denote the number of events in the time interval  $(0, t]$ .

(a) Define what it means that the process  $S_1, S_2, \ldots$  is a non-homogeneous Poisson process (NHPP) with intensity function  $w(t)$ . Define also the cumulative intensity function  $W(t)$  and write down the probability distribution of  $N(t)$  for a given  $t > 0$ .

A software system is subject to failures at random times caused by errors present in the code. Let  $N(t)$  be the cumulative number of failures experienced by time t. We

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assume that each failure is caused by exactly one error in the code, and that this error is successfully removed from the software after each failure. Hence  $N(t)$  also represents the cumulative number of errors detected and removed by time t.

A classical model for software reliability is the time dependent error detection model of Goel and Okumoto (the GO-model), where one assumes that  $N(t)$  is an NHPP with intensity function of the form

$$
w(t) = \alpha \beta e^{-\beta t}
$$

for parameters  $\alpha > 0, \beta > 0$ .

(b) Calculate the cumulative intensity function  $W(t)$ . What is the interpretation of  $W(t)$ ?

What is the limit of  $W(t)$  as  $t \to \infty$ ? Explain why this limit suggests an interpretation of the parameter  $\alpha$  as the "initial number of errors" in the code.

(c) Suppose that the software system has been run until time  $s > 0$ . The *conditional reliability function*  $R(t|s)$  of the software at time s is defined to be the probability that the software operates without failures for at least time  $t$  beyond time  $s$ . Show that

$$
R(t|s) = \exp(-e^{-\beta s}W(t)).
$$
\n(6)

The GO-model is in particular used in software testing. Suppose that values of  $\alpha$  and  $\beta$  are estimated from previous data and hence can be assumed known. Then (6) can be used to calculate an *optimal testing time*  $s_0$  by the requirement, on the testing time s, that the conditional reliability  $R(t_0|s)$  should be at least equal to a given value r  $(0 < r < 1)$ , for a given time  $t_0$ .

(d) Derive the corresponding expression for  $s_0$  in terms of  $\alpha, \beta, r$  and  $t_0$ . Discuss in particular the choice  $t_0 = \infty$ .

Recall the interpretation of the parameter  $\alpha$  as the "initial number of errors" in the code. In the *Jelinski-Moranda model for software reliability* (the JM-model) the *true* initial number of errors, a, is an unknown parameter, which is hence a non-negative integer which we shall assume is at least one.

The basic assumption of the JM-model is that the time  $T_i$  between failure number  $i-1$ and failure number i is exponentially distributed with hazard rate

$$
\lambda_i = (a - i + 1)b
$$

for  $i = 1, 2, \ldots, a$ , where  $b > 0$  is the second parameter of the model. It is further assumed that  $T_1, T_2, \ldots, T_a$  are stochastically independent.

(e) Explain why the time  $S_a$  of the ath failure can be interpreted as the time when all errors in the code are found.

Derive an exact expression for  $E(S_a)$ , i.e., the expected time when the software is errorfree. Then verify the approximation

$$
E(S_a) \approx \frac{\ln a}{b}
$$

when  $a$  is large.