Reliability of Fatigue-Prone Aircraft and Airline

Abstract

The limitation of the probability of any fatigue failure in a fleet of N fatigue-prone aircraft (FFPN) and fatigue failure rate (FFR) of airline (AL) is a problem of high priority. The offered solution of the problem is based on the acceptance full-scale fatigue test of an aircraft structure. If the result of this test is not good enough then this new type of aircraft will not be used in a service. Previously the redesign of this project should be done. For this strategy there are a maximum of FFPN and a maximum of FFR as functions of unknown parameters of a fatigue life distribution and of a model of fatigue crack growth. In this paper the approach is discussed which allows to limit these maximums for any unknown parameters of fatigue life and fatigue crack model. Numerical examples are given.

Keywords: Inspection program; Markov chain; Monte-Carlo; Reliability; p-set function; Weibull distributions

Introduction

This paper is in some way a review, a correction and a development of our previous publications devoted to an elimination of a fatigue failure of an aircraft (AC). We discuss here the economical effectiveness of an airline (AL) under a limitation of the fatigue failure rate (FFR) and the limitation of any fatigue failure probability in a fleet of N fatigue-prone aircraft (FFPN).

The study of the fatigue problem of the aircraft has a long history. The earliest reported accident was the wing failure of a Dornier Merkur on 23 september 1927 [1]. C Torkington [2] in his paper reminds: "In a two year period from 1942, about 20 Vickers Wellington bombers were lost in the UK as a result of fatigue failures of the wing main spar joints. In the war situation, if 20 failures in the UK were identified as fatigue, one can only guess that at least a similar number were lost over the sea or enemy territory".

The most significant accidents were the catastrophic failures of Comet (1953, 1954), Fokker F-27 (1968), F-111 (1969), Hawker Siddeley (1976) and Boeing - 707 - 321C (1977). The developed country like USA, introduced Fail-Safe Concept (FS) for fixed wing transport aircraft was later would serve as the framework for common international standards. This new standard resulted from the US. FAA Transport Category Airplane Fatigue Regulatory Review Conference held in March, 1977 [5]. The European position, primarily advocated by the United Kingdom, was that transport category aircraft should meet two standards, the fail-safe and the safe-life methods, for certification of fixed wing aircraft.

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The crash of three Comets was the most significant event, which has a very strong influence on the next aircraft airframe development. Special philosophy and system of aircraft development should have been created in order to prevent aircraft fatigue failure. The first main ideas of the system were offered just during the Comet inquiry [4] in October - November 1954. Much attention was paid to the scatter of fatigue life. This is the opinion of director of the Royal Aircraft Establishment: «... I would have given a safety life of about one ninth of what the specimen comes to, because the specimen might by chance have been the strongest...” [4]. The approach to the fatigue problem, which developed from these ideas, was called a Safe-Life (SL) approach.

Whereas, if you only work on a single specimen, you would have to give a safety life of about one tenth of what the specimen comes to, because the specimen might by chance have been the strongest...” [4]. The approach to the fatigue problem, which developed from these ideas, was called a Safe-Life (SL) approach. Basically this requires that all the parts of the structure, the failure of which could result in loss of the aircraft, are to be able to remain safely in use for a predetermined retirement life (specified life (SL)).

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The FS and the DT concepts make emphasis on the design and test. During fail-safe test we have to prove, that the requirements are met. But it should be taken into account that even if the requirements are met the “fail-safe” structure is not safe, if it is not timely inspected and repaired. “So far better title would be “inspection dependent”. This clearly puts the emphasis for safety on the inspector, and implies that, without inspection, things may well be dangerous” [7].

For the choice of the SL and the program of inspections for the FS and the SL methods correspondingly, several mathematical problems should be solved. The model of fatigue crack growth should be developed and the cumulative distribution function of fatigue life should be studied. The mathematics for the calculation of the probability of the fatigue crack detection should be developed also. There is a lot of publications devoted to these problems. We mention only the most significant ones.

Much attention to these problems was devoted in the papers of Yang JN, et al. [8-16]. The statistical crack growth model and the distribution on equivalent initial flaw size were studied in [14,15]. The models of fatigue crack are offered also in [17,18] and [16,19]. In two last papers the process of the fatigue crack growth is considered as random process.

Statistical estimation of economic life for aircraft structures and aircraft fleet maintenance based on structural reliability analysis was studied in [13,20,21]. Some state-of-the-art and new mathematical approach to the problem was presented in 8th IMA International Conference on Modelling in Industrial Maintenance and Reliability (MIMAR), Oxford, Institute of Mathematics and its Applications, 2014. In [22] the control limit policy for aging systems using Markov decision process, in [23] a cooperative game-based decision method for aircraft fleet maintenance are discussed. Minimax approach to the development of inspection program in order to provide the economical effectiveness of an airline (AL) under a limitation of the fatigue failure rate (FFR) and the limitation of any fatigue failure probability in a fleet of N fatigue-prone aircraft (FFPN) is offered in [24]. The theory of semi-Markov processes with rewards was used for the solution of these problems.

Usually, the reliability problem is considered as a problem of the probability theory when the cumulative distribution functions (CDFs) of corresponding random variables (a fatigue life, fatigue crack model parameters ...) are known already. But in this paper, which is the development of [24] and some solutions presented in [25], the main attention is devoted to the statistical problems when these CDFs are not known but the solution of the problem is based on the acceptance full-scale fatigue test of an aircraft structure. In Appendix A the planning of inspection intervals is considered as definition of some set of specific “prediction intervals” for “future observations” (the detection and the fatigue failure times of the fleet aircraft) based on the processing of the acceptance fatigue test. After this test one of the two decisions should be chosen: 1) to do the redesign of new type of AC if the result of the test is “too bad” or 2) make the development of inspection program in other case. In this case required reliability can be provided for any unknown parameter of fatigue crack growth. Special attention in this paper will be made also to the beginning of service of a fleet of a new type aircraft.

Nomination of Specified Life

There are two versions of the problem.

1. The nomination of the specified life as some value in the interval (0,∞).

2. The SL is already nominated but it is necessary to check the required reliability.

Estimate of specified life

Usually for fatigue life data processing both lognormal and Weibull distributions are used. If we use logarithm scale (if we use $X = \ln(T)$ instead of random variable (RV) $T$), both these distributions will become distributions with location and scale parameters

$$X = \theta_0 + \theta_1 T,$$

where $\theta_0, \theta_1$ are unknown parameters, RV $T$ has either cdf $F_X(x) = F(x)$, where $F(x)$ is cdf of standard normal or the standard smallest extreme value (sev), $F_X(x) = 1 - \exp(-\exp(x))$, distributions (for lognormal or Weibull distributions of $T$ correspondingly). In following we use logarithm scale.

Let $X = \{X_1,\ldots,X_n\}$ where $X_i, i = 1,\ldots,n$, is the fatigue life of AC (it is a random variable) in full-scale fatigue test, $Y = \{Y_1,\ldots,Y_N\}$ where $Y_j, j = 1,\ldots,N$, is the fatigue life (random variable) of AC in service. We consider the case when all components of vectors $X$ and $Y$ are independent RVs with the same cdf which has the location and scale parameters $F(X(x) = \Phi(x)$, where $\Phi(x)$ is cdf of standard normal or the standard smallest extreme value (sev).

In order to limit the probability of fatigue failure of any AC in service by small value $\varepsilon$ the SL, $\tau = \tau(X)$, should be found from the equation

$$\sup_{\theta_0} P(Z \leq \tau(X)) \leq \varepsilon. \quad (2)$$

For considered assumptions we have the following solution [25]

$$\hat{\tau} = \hat{\theta}_0 + t_{\hat{\theta}_1}, \quad (3)$$

where the random variables, estimates of $\theta_0, \theta_1$ (for example, estimates of Maximum Likelihood (ML) method) as function of $X = \{X_1, X_2,\ldots,X_n\}$ have the following structures

$$\hat{\theta}_0 = \hat{\theta}_0 + \hat{\theta}_1 \theta_0, \quad \hat{\theta}_1 = \hat{\theta}_1 \theta_1, \quad (4)$$

where $\theta_0$, $\theta_1$ are RVs corresponding to the estimates of $\theta_0, \theta_1$ using a sample $X = \{X_1,\ldots,X_n\}$ of the same size $n$ but when $\theta_0 = 0, \theta_1 = 1$; the RV $Z$ has the same type of CDF as $Z$ but $\theta_0 = 0, \theta_1 = 1$; $t$ is $\varepsilon$-quintile of RV $Y = (Z - \theta_0) / \theta_1$, $\varepsilon$ is allowed
probability of failure of any AC.

Fixed required specified life-test time limitation

Now we consider the case when the required SL, \( \tau \), is fixed already but using result of the acceptance full-scale fatigue test it is necessary to check the reliability requirement. And we consider also definition of the required test time.

Let us define

\[
\hat{\tau} = \hat{\theta}_0 + t_2 \hat{\theta}_1
\]

(5)

and we say that required reliability is provided if \( \hat{\tau} > \tau \).

The value \( t_2 \) we should define in such a way that probability of acceptance of the test and probability of fatigue failure of any AC in service should be limited for any parameter \( \theta \) by very small value, \( \varepsilon \)

\[
P(Z < \tau | \hat{\tau} > \tau) \leq \varepsilon.
\]

(7)

It can be shown [see [25]] that \( t_2 \) is the solution of the equation

\[
\sup_c F_{\hat{\theta}_0 + t_2 \hat{\theta}_1}(c) = \varepsilon
\]

(8)

where RV

\[
V_C = (c - \hat{\theta}_0) / \hat{\theta}_1.
\]

It is worth to note that if the test of all \( n \) ACs are made simultaneously, the defined by (4) structure of the parameter estimates takes place only if the test will be stopped at the moment of \( k \)th failure, \( k = 2, \ldots, n \). The value of \( k \) can be chosen taking into account some specific economical reason. Special attention should be made for the case when the scale parameter, \( \theta_1 \), is known. In this case it can be found the limitation of the test time without any failure. It can be shown that in this case the required SL, \( \tau \), can be excepted and required reliability \((1 - \varepsilon)\) will be provided if the smallest fatigue life of the tested AC will be more than

\[
x_{01} = \tau - t_2 \hat{\theta}_1
\]

where \( t_2 \) is the solution of the equation

\[
\max_c F_{\hat{\theta}_0 + t_2 \hat{\theta}_1}(c)(1 - F_{\hat{\theta}_0}(c - t_2 \hat{\theta}_1)) = \varepsilon.
\]

(9)

Numerical examples of the test time nomination

Simultaneously fatigue tests of 6 airframes of the same type have been made but only up to 4th fatigue failure. So we know only 4 first minimal fatigue lives: \( (t_1, \ldots, t_4) = (59971, 72600; 77630; 80863) \) and correspondingly for \( x_i = \ln(t_i) \), \( i = 1,\ldots, 4 \), we know \( x = (x_1, \ldots, x_{10}) = (11.002; 11.193; 11.260; 11.3005) \). There are 100 aircraft in operation and there is a requirement that the probability of at least one fatigue failure before \( t_{\text{SL}} = 50000 \) cycles should not exceed \( p = 0.05 \). Then \( \tau = \log(50000) = 10.82 \). We can be sure of the required reliability if \( \hat{\theta}_0 + t_2 \hat{\theta}_1 > \tau \).

Let us consider the lognormal distribution of the fatigue life. Then using the Lifereg procedure of SAS system we can easily get ML estimates \( \hat{\theta}_0 = 11.26, \hat{\theta}_1 = 0.145 \). And then using Monte Carlo method to get the cdf of RV \( V_C \) (5000 samples) we have, that \( t_2 = -7.055 \) is the root of the equation

\[
\max(1 - (1 - \Phi(c))^n) F_{\hat{\theta}_0}(c) = p = 0.05
\]

(10)

Where

\[
V_C = (c - \hat{\theta}_0) / \hat{\theta}_1.
\]

Accordingly \( \hat{\tau} = \hat{\theta}_0 + t_2 \hat{\theta}_1 = 11.26 - 7.055 \times 0.145 = 10.237 \). This value is less than required \( \tau = 10.82 \). So the required reliability is not provided. Now let us consider the case when \( \hat{\theta}_0 = 0.346 \) is known and a new fatigue test after some structure retrofit has to be made. And we have to know the time limit of the fatigue test without any failure, which will be enough to be sure of the required reliability. In this case \( t_2 = -2.04 \) is the root of the equation

\[
\max(1 - (1 - \Phi(c))^n)(1 - \Phi(c - t)\bar{c}) = 0.05, \quad N = 100; n = 6.
\]

(11)

So for the required time limit of the fatigue test without the failure (in logarithm scale) we have

\[
x_{\text{WF}} = x_{10} = \tau - t_2 \hat{\theta}_1 = 10.82 - (-2.0425) \times 0.346 = 11.52648
\]

or in the natural scale

\[
t_{\text{WF}} = \exp(11.52648) = 101365.
\]

It is worth to note that for the same initial data but for Weibull distributions of fatigue live the required time of test without failure is equal to 179836.

Developing of Inspection Program

Limitation of fatigue failure rate

Now we suppose that reliability of every AC is defined by fatigue crack which is discovered with probability 1 if inspection is made in the interval \( (T_{i0}, T_i) \) where \( T_{i0}, T_i \) are random variables: \( T_{i0} \) is the time when the fatigue crack can be detected, \( T_i \) the time when the fatigue failure takes place. Here we consider the simplest case: interval between the inspections is equal to the constant \( t_{\text{SL}} / (n + 1) \), where \( t_{\text{SL}} \) is the aircraft specified life (SL) (the retirement time). The process of an operation of AL can be considered as a Markov chain (MCh) [25] with \( n + 4 \) states. The states \( E_1, E_2, \ldots, E_{n+1} \) correspond to an AC operation in the time intervals \( (t_0, t_1), (t_1, t_2), \ldots, (t_{n+1}) \), \( t_0 = 0 \), \( t_{n+1} = t_{\text{SL}} \). The states \( E_{n+2}, E_{n+3}, \) and \( E_{n+4} \) correspond to the events: AC reaches the SL without any problem, the fatigue failure (FF) or the fatigue crack detection (CD) take place and in all these three cases the new AC

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is purchased and begins the service in first interval.

In the corresponding transition probability matrix, \( P_{\text{AL}} \), let \( v_i \) be the probability of a crack detection during the inspection number \( i \), let \( q_i \) be the probability of the failure in service time interval \( (t_{i-1}, t_i] \), and let \( u_i = 1 - v_i - q_i \) be the probability of successful transition to the next state. In our model we also assume that an aircraft is discarded from a service at \( t_{n+1} \) even if there is no any crack discovered by inspection at the time moment \( t_{n+1} \).

The inspection at the end of \( (n+1)\)-th interval (in state \( E_{n+1} \)) does not change the reliability but it is made in order to know the state of aircraft (whether there is a fatigue crack or there is no fatigue crack). It can be shown that

\[
\prod_{i=1}^{n+1} (1 - u_i - b_i q_i) = \prod_{i=1}^{n+1} \left( \sum_{j=1}^{n+1} q_{ij} \right) = \prod_{i=1}^{n+1} \left( \sum_{j=1}^{n+1} a_{ij} \right) = \prod_{i=1}^{n+1} \left( \sum_{j=1}^{n+1} c_{ij} \right) = \prod_{i=1}^{n+1} \left( \sum_{j=1}^{n+1} d_{ij} \right) = \prod_{i=1}^{n+1} \left( \sum_{j=1}^{n+1} e_{ij} \right) = \prod_{i=1}^{n+1} \left( \sum_{j=1}^{n+1} f_{ij} \right).
\]

Table 1: Matrix of transition probabilities \( P_{\text{AL}} \).

| \( E_1 \) | \( E_2 \) | \( E_3 \) | \( \ldots \) | \( E_{n+1} \) | \( E_{n+2} \) (SL) | \( E_{n+2} \) (FF) | \( E_{n+2} \) (CD) |
|---------|---------|---------|----------|---------|----------------|----------------|----------------|
| \( E_1 \) | 0       | \( u_1 \) | 0        | \( \ldots \) | 0       | 0               | 0               | 0               |
| \( E_2 \) | 0       | 0       | \( u_2 \) | \( \ldots \) | 0       | 0               | 0               | 0               |
| \( E_3 \) | 0       | 0       | 0        | \( \ldots \) | 0       | 0               | 0               | 0               |
| \( \ldots \) | \( \ldots \) | \( \ldots \) | \( \ldots \) | \( \ldots \) | \( \ldots \) | \( \ldots \) | \( \ldots \) |
| \( E_{n-1} \) | 0       | 0       | 0        | \( \ldots \) | 0       | 0               | 0               | 0               |
| \( E_n \) | 0       | 0       | 0        | \( \ldots \) | 0       | 0               | 0               | 0               |
| \( E_{n+1} \) | 0       | 0       | 0        | \( \ldots \) | 0       | 0               | 0               | 0               |

Using the theory of the semi-Markov process with rewards and the definition of the matrix \( P_{\text{AL}} \), we can get the vector of stationary probabilities, \( \pi = (\pi_1, \ldots, \pi_{n+1}) \), which is defined by the equation system:

\[
\pi P_{\text{AL}} = \pi, \quad \sum_{i=1}^{n+1} \pi_i = 1
\]  

(12)

and the airline gain

\[
g(n) = \sum_{i=1}^{n+1} \pi_i G_i(n) \]  

(12a)

where

\[
g_i(n) = \begin{cases} a_i u_i + b_i q_i + c_i v_i, & i = 1, \ldots, n + 1, \\ d_i, & i = n + 2, \ldots, n + 4, \end{cases}
\]

\( a_i \) is the reward defined by the successful transition from one operation interval to the following one and the cost of one inspection; \( b_i, c_i \) and \( d_i \) correspond to transition to the states \( E_{n+3} \) (FF), \( E_{n+4} \) (CD) and then to the state \( E_n \) (the "cost" of FF of AC, fatigue crack detection, acquisition of new AC). Let us note that if \( a_i = t_i - t_{i-1} \), \( b = c = d = 0 \) then

\[
g(n) = \sum_{i=1}^{n+4} \pi_i G_i(n) = \sum_{i=1}^{n+4} \pi_i (t_i - t_{i-1}) \]  

(13)

Let the cdf of the vector \( (T_j, T_L) \) and the matrix \( P_{\text{AL}} \) are defined by the known parameter \( \theta \). Then we should consider \( g(n, \theta) \) and \( G_\theta(n, \theta) \). The value \( L_j = g(n, \theta) / \pi_j \) defines the mean time to return to the same state \( jE \). So \( \lambda_j(n, \theta) = 1 / L_j \) is the FFR.

If \( \theta \) is known we calculate the gain as a function of \( n, g(n, \theta) \), and choose the number \( n_j \) corresponding to the maximum of gain:

\[
n_j(\theta) = \arg \max g(n, \theta).
\]

Then we calculate FFR as function of \( n, \lambda_j(n, \theta) \), and choose \( n_j \) in such a way that for any \( n \geq n_j \) the function \( \lambda_j(n, \theta) \) will be equal or less than some value \( \lambda : n_j(\lambda, \theta) = \min \{ n : \lambda_j(n, \theta) \leq \lambda, \ \text{for all} \ n \geq n_j(\lambda, \theta) \} \). And finally \( n = n_j(\lambda, \theta) = \max(n_j, n_j(\lambda, \theta)) \), Figure 1.

But if \( \theta \) is not known the value \( n \) is a RV \( \hat{n} = n_j(\lambda, \hat{\theta}) \). For the approximate solution the confidence interval for the \( \theta \) can
be used. But some uncertainty appears: confidence level is not defined by required reliability. The precise solution of the limiting value \( \lambda \) can be found in case if the fatigue test is acceptance test. The result of acceptance test can be used to calculate the estimate of the parameter \( \theta, \hat{\theta} = \hat{\theta}(x_1, \ldots, x_N) \). The service of a new type of aircraft will not take place if the result of the fatigue test in a laboratory is “too bad” (previously, the redesign of the new type of AC should be made). We say that in this case the event \( \hat{\theta} \not\in \Theta_0 \) takes place, where \( \Theta_0 \subseteq \Theta \) is a parameter space (for example, \( \hat{\theta} \not\in \Theta_0 \) if the test fatigue life, \( t_v \), is lower than some limit; or \( \frac{n_{\hat{\lambda}}(\lambda, \hat{\theta})}{\lambda} \) is too large,...).

Let \( \lambda_{FD} \) be a preliminary designed allowed FFR which is the solution of the equation

\[
\sup_{\lambda} E(\hat{\lambda}_F(n(\lambda_{FD}, \hat{\theta}), \theta)) | \hat{\theta} \in \Theta_0)P(\hat{\theta} \in \Theta_0) = \lambda \tag{14}
\]

where \( E(\lambda) \) is the expectation corresponding to the distribution of \( \hat{\theta} \) under condition \( \hat{\theta} \in \Theta_0 \).

Then the number of inspections \( n_{\hat{\lambda}}(\lambda_{FD}, \hat{\theta}) \) provides the limitation of \( \lambda \) independently of any unknown parameter \( \theta \).

**Limitation of probability of any fatigue failure in fleet of aircraft**

Now we consider the case when the operation of all \( N \) aircraft will be stopped if any fatigue crack will be detected. So in order to limit the probability of fatigue failure in the fleet (FFPN) it is enough to find at least one fatigue crack before the failure of any aircraft in the fleet takes place. Let \( t_{i0}^+, t_{i1}^+, \ldots, t_{ik}^+ \) to be “calendar” time moment when \( k \)th aircraft begins the service, \( T_{i_k} = t_{i0}^+, T_{i_k} = t_{i0}^+ + t_{i1}^+, \ldots, k = 1, 2, \ldots, N \), to be the random calendar time moments when fatigue crack can be discovered and fatigue failure of AC takes place correspondingly, see Figure 2. And let \( K_{SL} = \{ k : T_{i_k} < t_{SL}, k = 1, 2, \ldots, N \} \) be a set of indexes of aircraft the failure of which can take a place if an inspection will not take the place, \( T_{f}^+ = \min\{T_{i_k}^+ : k \in K_{SL}\} \), is a calendar time corresponding the first failure in the fleet without inspections and let \( K_f = \{ k : 1 \leq k \leq N, \min(t_{i_k}^+, T_{f}^+) \} \) is the set of aircraft the service of which begin before the first failure in the fleet, \( K_{SLF} = K_{SL} \cap K_f \); and let us define:

\[
T_{SLF}^+ = \min\{T_{i_k}^+, T_{f}^+\}, \quad T_{SLF}^+ = \min\{T_{i_k}^+, T_{f}^+\},
\]

\[
R_{SLF} = \max\left\{ j : t_{i_j}^+ < T_{SLF}^+, j = 0, 1, 2, \ldots \right\},
\]

\[
R_{SLF} = \max\left\{ j : t_{i_j}^+ < T_{SLF}^+, j = 0, 1, 2, \ldots \right\}, \quad R_{SLF} = R_{SLF} - R_{SLF}
\]

\( k \in K_{SLF} \).

For fixed value of parameter \( \theta \) the probability of any failure in the fleet \( p_{NNW}(n, \theta) \) will be equal to expected value of random variable \( P_{NNW}(n, \theta) \)

\[
p_{NNW}(n, \theta) = E^T(P_{NNW}(n, \theta)), \quad \text{where}
\]

\[
P_{NNW}(n, \theta) = \begin{cases} P_{1WW}, & \text{if } w = 1, \\ (1-w)^\theta, & \text{if } 0 \leq w < 1, \\ 0, & \text{if } w \geq 1 \end{cases}, \quad P_{1WW} = \begin{cases} 1, & \text{if } R_f = 0, \\ 0, & \text{if } R_f \geq 1 \end{cases}
\]

\( w \) is a human factor: a probability, that the planned inspection will be made, \( R_f = \sum_{k=1}^{K_{SLF}} R_{ik} \) is the total random number of inspections before the first failure in the whole fleet; \( E^T(.) \) is the expectation corresponding to the distribution of a set of vectors ((\( T_{Dk}^+, T_{Ck}^+), k = 1, 2, \ldots N \)). Let us note that if \( w = 1 \) but the expected probability is very small then approximately

\[
E(P_{NNW}(n, \theta)) = E(1 - \prod_{k=1}^{N} (1 - P_{1WW})) \approx E(1 - \prod_{k=1}^{N} P_{1WW}). \quad (15a)
\]

The necessary calculation can be made by the use of Monte Carlo method taking into account the distribution of random variables \( T_i \) and \( T_f \). If parameter \( \theta \) is known then the number of the inspections, \( n(p, \theta) \), required to limit the FFPN by a value \( p \) is defined by the equation

\[
n(p, \theta) = \min(r : p_{NNW}(n, \theta) \leq p \quad \text{for all } r > n(p, \theta), \quad r = 1, 2, \ldots ).
\]

(16)

For the case of using the result of the acceptance test for the estimation of the parameter \( \theta \) the required limitation of the FFPN by a value \( p \) is provided for any unknown \( \theta \) if the number
of the inspections will be defined by value \( n(p_{FD}, \hat{\theta}) \), where \( p_{FD} \) is the solution of the equation

\[
\sup_{\theta} E^\theta \{E^T (P_{NW} (n(p_{FD}, \hat{\theta})) | \hat{\theta} \in \Theta_0) \} P(\hat{\theta} \in \Theta_0) = p
\]  

(17)

where again \( E^T (.) \) is the expectation corresponding to the distribution of a set of vectors \((T_{d_i}^+, T_{d_i}^-)\), \( k = 1, \ldots, N \), the \( E^\theta \) is the expectation corresponding to the distribution of \( \hat{\theta} \) under condition \( \hat{\theta} \in \Theta_0 \).

**Numerical Example**

We suppose that the equation \( a(t) = a \exp(\alpha t) \), where \( a \) is some constant, \( Q \) is some RV, describes the development of fatigue crack in the interval, \((t_x, t_y)\) where \( t_x \) is a time when the crack becomes detectable \( (a(t_x) = a_x) \) and \( t_y \) is a time when the crack reaches its critical size \( (a(t_y) = a_c) \) and fatigue failure takes place. Corresponding random variables are defined by equations:

\[
T_x = (\log a_x - \log a) / \log C_x / \log Q, \quad T_y = (\log a_c - \log a) / \log C_c / \log Q.
\]

Let us denote \( X = \log Q \) and \( Y = \log C_x \), where \( C_x = \log a_x - \log a \). Here we assume that \( C_x \) is some constant. From the analysis of the fatigue test data it can be assumed, that the \( \log T_x = \log C_x - \log Q \) is distributed normally. It can take place only if \( X = \log Q \) has normal distribution. Additionally we assume that standard deviation of \( \log(Q) \) is equal to 0.346.

Now we consider an example of the problem to limit the FFPN by the value \( p = 0.05 \). Suppose that during fatigue test we see the fatigue crack (see Figure 2.22 in [1]) and get the following data:

\[
\log(Q) = -8.588; \quad \alpha = 0.286\text{mm}.
\]

It is known: \( \alpha_x = 20\text{mm}, \alpha_c = 237\text{mm} \) [1]. Assume that \( t_y = 40000\text{h}, w = 0.9 \); there is 10 aircraft in the fleet, the interval between the aircraft putting into operation as \( t_{i+1} - t_i = 500\text{h} \); required reliability \( R = 0.95 \), allowed failure probability \( \varepsilon = 1 - R = 0.05 \), a number of allowed maximal inspections is equal to 20 (the redesign of aircraft should be made if required reliability \( R = 0.95 \) is provided only for the inspection number, \( n = n(\varepsilon, \hat{\theta}) \), more than 20; this requirement defines the set \( \Theta_0 \).

The calculations made by the use of Monte Carlo method show: if \( \hat{\theta}_0 = \log(Q) = -8.588, \hat{\theta}_1 = 0.0346 \) are fixed, all inspection intervals are equal, then it can be calculated that 9 inspections for each aircraft during the operating time should be carry out to ensure the required reliability. But if it taken into account that \( \hat{\theta}_0 = \log(Q) = -8.588 \) is only the estimate of the unknown parameter then 16 inspections should be chosen corresponding to \( p_{FD} = 0.01 \). We see that in order to limit the fleet failure probability for any unknown \( \theta_0 \) by the value \( \varepsilon = 0.05 \) the inspection interval should be calculated using \( p_{FD} = 0.01 \).

**Appendix A**

Here we consider only the limitation of any fatigue failure in a fleet of N aircraft and show a connection of the development of the inspection program with the definition of a prediction interval for future observation [26].

Let \( Z = (Z_1, \ldots, Z_N, \ldots) \), where \( Z^+_k = (T^{+}_{d,k}, T^{-}_{d,k}) \), \( k = 1, \ldots, N \) are some previously defined vectors, the vector \( X \) defines the result of the acceptance full-scale fatigue test of an aircraft structure. It is supposed that the class is known \( \{P_v, \theta \in \Theta\} \) to which the probability distribution of the random vector \( W=(Z, X) \) is assumed to belong. Of the parameter \( \theta \), which labels the distribution, it is presumably known only that it lies in a certain set \( \Theta \), the parameter space and \( \hat{\theta} \) is the estimate of the parameter \( \theta \) as a result of processing \( X \) It is useful to note that the choice of the program with the \( n \) inspections defines some random set function

\[
S^+(\hat{\theta}, \Theta_0, n) = \bigcup_{i \leq N} S^+_i(\hat{\theta}, \Theta_0, n)
\]

(A1)

where

\[
S^+_i(\hat{\theta}, \Theta_0, n) = \begin{cases} 
\bigcup_{j \leq N} S^{+}_{j,\hat{\theta}}(n), & \text{if } \hat{\theta} \in \Theta_0, \\
\emptyset, & \text{if } \hat{\theta} \notin \Theta_0.
\end{cases}
\]

A function \( S^+(\hat{\theta}, \Theta_0, n) \) is shown in Figure 3.

**Figure 3:** Example of \( S^+(\hat{\theta}, \Theta_0, n) \).

This function defines some set of specific “prediction intervals” for the set of the vectors \( \{Z^+_k, k = 1, \ldots, N\} \) based on the observation of the \( X \) (or based on the estimate \( \hat{\theta} \) of the parameter.
which defines the probability of any fatigue failure in the fleet, FFPN. We remind that if X is the result of the acceptance test the required limitation of the FFPN by a value \( p \) is provided for any unknown \( \theta \) if the number of the inspections will be defined by value \( n(p_{FD}, \hat{\theta}) \), where [we remind] \( p_{FD} \) is the solution of the equation

\[
\sup_{\hat{\theta}} E^T \left( P_{P_{NW}} \left( n(p_{FD}, \hat{\theta}) \right) | \hat{\theta} \in \Theta_0 \right) P(\hat{\theta} \in \Theta_0) = p \quad (A2)
\]

where again \( E^T(.) \) is the expectation corresponding to the distribution of a set of vectors \( (T^*_{n0}, T^*_{n1}, \ldots, T^*_{nN}) \), the \( E^\hat{\theta}(.) \) is the expectation corresponding to the distribution of \( \hat{\theta} \) under

\[
\sup_{\hat{n}} \left( 1 - \prod_{k=1}^{N} \left( 1 - \sum_{j \leq R} P(Z^*_{ij} \in S^*_{ij}(n(p_{FD}, \hat{\theta})) \right) | \hat{\theta} \in \Theta_0 \right) P(\hat{\theta} \in \Theta_0) = p \quad (A4)
\]

If the left side part of (A4) is small enough then instead of (A4) the equation (A5) can be used

\[
\sup_{\hat{n}} \left( 1 - \sum_{k=1}^{N} \sum_{j \leq R} P(Z^*_{ij} \in S^*_{ij}(n(p_{FD}, \hat{\theta})) | \hat{\theta} \in \Theta_0 \right) P(\hat{\theta} \in \Theta_0) = p \quad (A5)
\]

Note: It should be noted that in [27] similar problem was considered but with some mistakes. The equation (A2) is not given at all but the equation (A5) is given without explanation that it can be used only if \( p \) is very small.

Equations (A2 – A5) show that the FFPN will be limited by the value \( p \) for any unknown parameter \( \theta \).

Conclusion

Some mathematics for the solution of the problem of the limitation of the probability of any fatigue failure in a fleet of N fatigue-prone aircraft (FFPN) and fatigue failure rate (FFR) of some airline is offered. Usually, the reliability problem is considered as a problem of the theory of probability when the cumulative distribution functions (CDFs) of the corresponding random variables (the fatigue life, the fatigue crack model parameters ...) are known already. But in this paper the main attention is devoted to the statistical problem when these CDFs are not known but the solution of the problem is based on the acceptance full-scale fatigue test of an aircraft structure. In the Appendix A the planning of inspection intervals is considered as definition of some set of specific “prediction intervals” for “the future observations” (the detection and the fatigue failure times of the fleet aircraft). They are based on the processing of the acceptance fatigue test. After this test one of the two decisions should be chosen: 1) do the redesign of new type of AC if the result of the test is “too bad” or 2) to make the choice of the number of inspection, \( n = n(\theta, p_{FD}) \), as function of \( \hat{\theta} \) and specific \( p_{FD} \), defined in this paper. In this case required reliability can be provided for any unknown parameter of fatigue crack growth.

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Conflict of Interest

The authors declare that there is no conflict of interests regarding the publication of this paper.

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