Reliability of Aircraft Structure Joints Under Corrosion-Fatigue Damage

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Presentation Content:

- 1. Reliability Analysis for Aging Structures
- 2. Corrosion-Fatigue Damage Modeling
- 3. Stochastic Multi-Scale FE Analysis
- 4. Aircraft Lapjoint Example
- 5. Concluding Remarks

1. Reliability Analysis for Aging Components



Risk-Based Maintenance Analysis Concept



Stochastic Aircraft Operating Environment



2. Corrosion-Fatigue Damage Modeling

Six Aging Stages (A-F)



Model 1: Simultaneous CF Model (SCF)

Crack Initiation

Strain-Life Curve for 2024-T3 Sheet

Crack Propagation





$$\mathbf{a}(t) = \mathrm{IDS} + [\sum_{n,t} \mathbf{a}(n)^{\alpha} + \mathbf{p}(t)^{\beta}]^{\gamma}$$



Forman FCG Model (NASA JPL, 1996)

$$\frac{da}{dN} = \frac{C(1-R)^{m}\Delta K^{n}(\Delta K - \Delta K_{th})^{p}}{[(1-R)K_{c} - \Delta K)]^{q}}$$
$$\Delta K_{CF} = \psi(t)\Delta K_{F}$$
Uncertain Parameters
$$\psi(t) = \sqrt{1 + \frac{p(t)}{a(n)}}$$



Pitting Corrosion-Fatigue Model (FASTRAN-CF)

$$\Delta K'_{eff} = \psi(t) \Delta K_{eff} \qquad \psi(t) = \sqrt{1 + \frac{a_{pit}(t)}{c_{crack}(t)}} - FASTRAN-CF$$

Model 2: Wei Corrosion-Fatigue Model (WCF)

DEFECT SIZE



Pit Growth

$$\mathbf{a} = \left\{ \left[\frac{3\mathrm{MI}_{\mathrm{p0}}}{2\pi\mathrm{nF}\rho} \exp\left(-\frac{\Delta\mathrm{H}}{\mathrm{RT}}\right) \right] \mathbf{t} + a_0^3 \right\}^{1/2}$$

Corrosion-Fatigue Life Model

$$\mathbf{t}_{\mathrm{f}} = \mathbf{t}_{\mathrm{ci}} + \mathbf{t}_{\mathrm{tc}} + \mathbf{t}_{\mathrm{cg}}$$

$$t_{ci} = \frac{2\pi n F \rho}{3MI_{P0}} (a_{ci}^3 - a_0^3) \exp\left(\frac{\Delta H}{RT}\right)$$

$$t_{tc} = \frac{2(\sqrt{\pi})^{n_{C}} \left[(\sqrt{a_{ci}})^{2-n_{C}} - (\sqrt{a_{tc}})^{2-n_{C}} \right]}{\nu(n_{C} - 2)C_{C}(2.2K_{t}\Delta\sigma)^{n_{C}}}$$

$$t_{cg} = \int_{a_{tc}}^{a_{f}} \frac{1}{\upsilon C_{c} \left(\Delta \sigma \sqrt{\pi} \right)^{n_{C}}} \left(\frac{0.324r_{0} + a}{1.086r_{0} \sqrt{a} + 0.681 \left(\sqrt{a} \right)^{3}} \right)^{n_{C}} da$$

Comparative Results





Comparative Results for SCF Model (ProCORFA) vs. CCCF Model (FASTRAN-CF) and WCF Model (Wei)



Stochastic Corrosion-Fatigue Life Simulation





3. Stochastic Multi-Scale FE Analysis



Local Model:

- Solid elements
- Includes contacts
- Obtain BCs from global model
- Consider stochastic parameters
- Detailed local stresses

Global Model:

- Shell and beam elements
- "Weld" stringers and frames
 with skin panel
- Linear analysis
- Find critical locations

Very Local Model:

- Axisymmetric elements
- Material and full contact nonlinearity
- Residual stress and interference analysis

Stochastic Modeling for Aircraft Lap Joint

Define Stochastic Variables

Loading/Cabin Pressure Cycles Manufacturing Tolerances Material (Elastic modulus, etc.) Corrosion Material Loss

Governing Stochastic Variables

Rivet interference Skin panel thickness Rivet hole location Rivet hole diameters Corrosion Loss



Loading spectrum



Manufacturing tolerances

Remarks: We consider up to 55 random variables within a single stochastic FEA. However, only a limited number of variables, say 5-6, are of significance at different critical locations

Stochastic Response Surface Approximations

Second-Order (SO) Approximation of Stochastic Fields

Explicit Formulation: Using function approximation via nonlinear regression



y



---- Least-square fitting (y is explicit)

Limited

Refined

Convergence: Minimizing Mean-Square Error (in Mean-Square sense) Causal relationship. We used spatial Krigging.

High-Order (HO) Approximation of Stochastic Fields

Implicit Formulation: Using joint PDF estimation of $z = [y, x]^T$

$$\int_{\mathbf{x}_{-j}}^{\mathbf{x}_{-j}} \frac{f(y|\mathbf{x}) = \frac{f(y,\mathbf{x})}{f(\mathbf{x})}}{E[y|\mathbf{x}] = \int_{0}^{\infty} yf(y|\mathbf{x})dy}$$

--- Stochastic Neural Networks (y is implicit) Decomposes overall complex JPDF in localized simple JPDFs. Solution is obtained by stochastic interpolation

Convergence: Using Maximumum Likelihood Function (in Probability sense) Non-causal relationship. We use 2-and 3-Level Hierarchical Models

HO SF Models Applicable to Highly Non-Gaussian RS



Two-Level Hierarchical Model Versus Krigging with 11% Noise



4. Aircraft Lapjoint Example



Inside Aircraft

Load Transfer in the Lapjoint

Sensitivity Studies for Identifying Critical Stress Locations and Input Variables

Random Variables used in Sensitivity Study

Elasticity modulus Thickness change of the splices Location shift of rivets Diameters of rivet holes Interferences at the rivet holes

Governing Random Variables

Thickness change of splices Hole location shifts Interference between the rivet and holes



Comparison of FEA and Response Surface Results - Location 7 of Hole 3



Direct Stochastic FE Analysis

Mean Stresses from Direct FE Analysis (ksi)





Response Surfaces

Stochastic Response Surface Sections in 2D

Response of L7 at Hole 3

Input Variables:

Interference of Hole 1 Thickness change of Plate #1

Fixed values for other variables

Remark: Both variables show significant influences to the response surface





Stress Range at L7 of Hole 3

Mean Stress at L7 of Hole 3

Bivariate Stress Distribution at L3 of Hole 1









Simulation of Corrosion Surface Topography





Simulation of FE Model with Material Loss Due to Corrosion (Details of Local Lap Joint Model)



Histogram Comparisons for With and Without Material Loss Due to Corrosion (at L7 of Hole #3)



No Corrosion





With Corrosion

Stress ranges

Mean Stresses





5. Concluding Remarks

1. Computational reliability analysis for aging aircraft structures using physicsbased stochastic corrosion-fatigue damage models provides a quantitative tool for reducing maintenance while maintaining the safety.

Availability of statistical data and their collection and use based on an in-depth understanding of the physics and stochastic modeling of the problem represent critical aspects of probabilistic technology implementation (need of sharing the available data between the people with data and the people with physics and stochastic modeling expertise that are often in different business organizations... cross communication is key!).

- 2. Airport rotation is an important factor for lowering aircraft risks per fleet.
- 3. IDS sizes near surface, corrosion pit depths, airport environment severity, number of flights/day, FE modeling including manufacturing tolerances, and inspection quality appears to be the main critical factors for aircraft risk prediction.
- 4. Stochastic FEA accuracy has a significant impact on aircraft risk prediction. Neglecting structural uncertainties is highly unconservative.