

The 86th Annual Meeting of the TRB

Development of Fatigue Cracking Performance Prediction Models for Flexible Pavements Using LTPP Database

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I. Introduction

Background and Objectives

- ◆ Predictive models used in pavement design, evaluation, rehabilitation, & management activities
- ◆ Evolves from purely empirical toward mechanistic-empirical approaches in the proposed MEPDG (DG2002)
- ◆ Focus on **predicting fatigue cracking of AC pavements using the LTPP database (www.datapave.com)**



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Outline

- ◆ I. Introduction
- ◆ II. Review of Existing Models
- ◆ III. Database Preparation
- ◆ IV. Analysis of Existing Models
- ◆ V. Development of Tentative Fatigue Cracking Models
- ◆ VI. Concluding Remarks



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II. Review of Existing Models

- ◆ 1. Various models for predicting N_f

$$N_f = k_1(\varepsilon_i)^{-k_2} |E^*|^{-k_3}$$

Organization (Year)	k ₁	k ₂	k ₃
Asphalt Institute (1981)	0.0796	3.291	0.854
Shell Oil (1982)	0.0685	5.671	2.363
Belgian Road Research Center (1984)	4.92x10 ⁻¹⁴	4.76	0
UC-Berkeley (1984)	0.0636	3.291	0.854
Transport and Road Research Laboratory (1984)	1.66x10 ⁻¹⁰	4.32	0
Illinois (1987)	5x10 ⁻⁶	3.0	0
U.S. Army (1988)	478.63	5.0	2.66
Indian (1999)	0.1001	3.565	1.474
Mn/ROAD (2003)	2.83	3.21	0



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Review of Existing Models

New Concepts of the Proposed MEPDG

Mechanistic-Empirical Pavement Design Guide (NCHRP 1-37A)

- ◆ 2. the revised MS-1 fatigue cracking model

$$N_f = 0.00432 * \beta_{f1} * C * \left(\frac{1}{\varepsilon_t} \right)^{3.291^{\beta_{f2}}} \left(\frac{1}{E} \right)^{0.854^{\beta_{f3}}}$$

$$C = 10^{4.84^{\beta_f} \left(\frac{V_p}{V_a + V_b} - 0.69 \right)}$$

$$F.C. = \left(\frac{6000}{1 + e^{(C_1^* C_f + C_2^* C_s^* \log(D_f^* 100))}} \right) * \left(\frac{1}{60} \right)$$

- ◆ Cumulative fatigue damage (D_f) using more complex Axle Load Spectra (ALS) concept

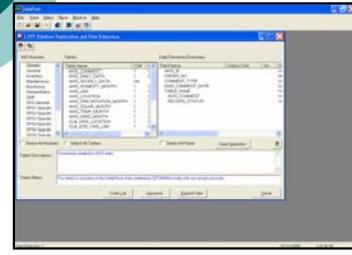
$$D_f = \sum_{i=1}^k \frac{n_i}{N_{fi}}$$

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III. Database Preparation

1. Long-Term Pavement Performance (LTPP) Program

- ◆ GPS-1 (Granular Base) & GPS-2 (Bound Base)



DataPave 3.0



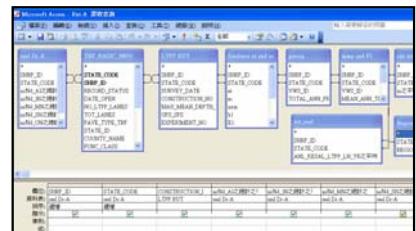
DataPave Online

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Database Preparation

2. Retrieval of Required Data

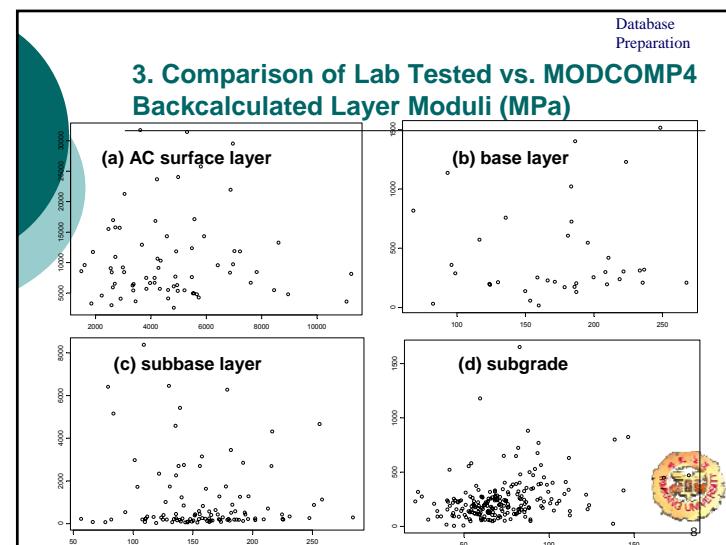
- ◆ IMS Modules/Tables/Data Elements:
 - Climatic
 - Maintenance
 - Testing
 - General
 - Monitoring
 - Traffic
 - Inventory
 - Rehabilitation



Microsoft Access

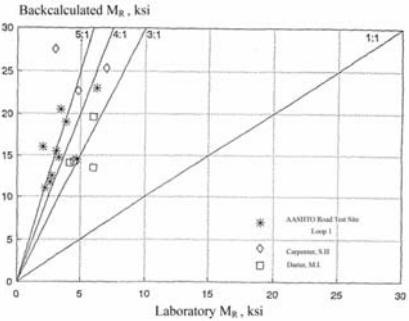
Existing models 10~15 items, DG2002 45~50 items
Batch BISAR Program Runs: Tensile strain of the AC layer

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Database
Preparation

Check with AASHTO 1993's Recommendation



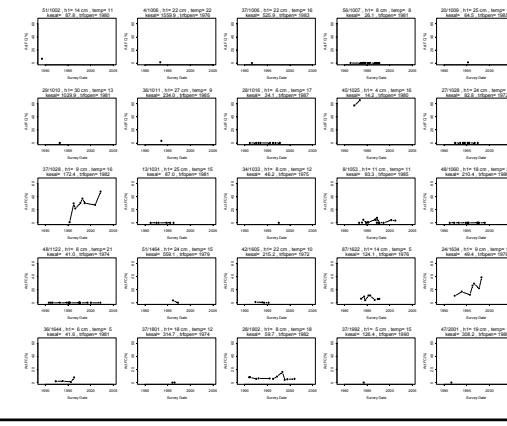
Subgrade: Consistent with an adjustment factor of 0.33
(Also note high variations of backcalculated AC surface, Base, and Subbase Moduli on previous slide)



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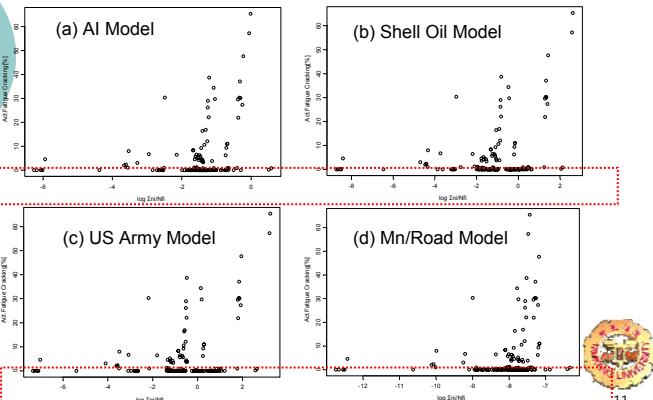
Database
Preparation

4. Graphical Representation and Data Cleaning



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IV. Analysis of Existing Models



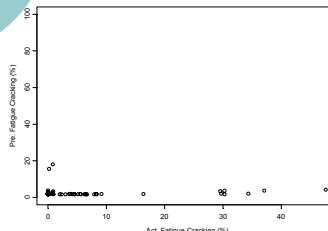
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Analysis of
Existing Models

Goodness of Fit of the Existing Models

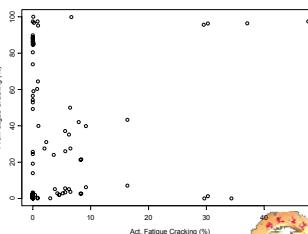
(a) AI Model and Ali & Tayabji Model

$$\% \text{FatigueCracking} = \frac{0.021}{0.027 + e^{(-0.057D_f)}}$$



$R^2=0.0065$, SEE=7.881, n=140

(b) MEPDG Models
Using DG2002 Program



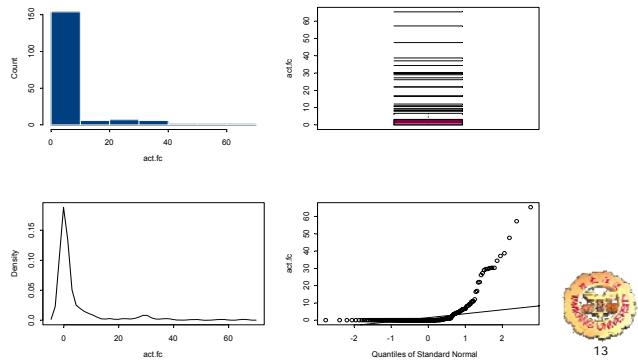
$R^2=0.0391$, SEE=7.751, n=140



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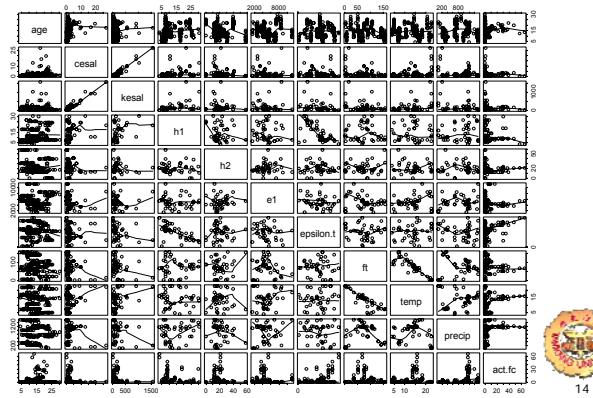
V. Development of Tentative Fatigue Cracking Models

1. Preliminary Analysis (Univariate Data Analysis)



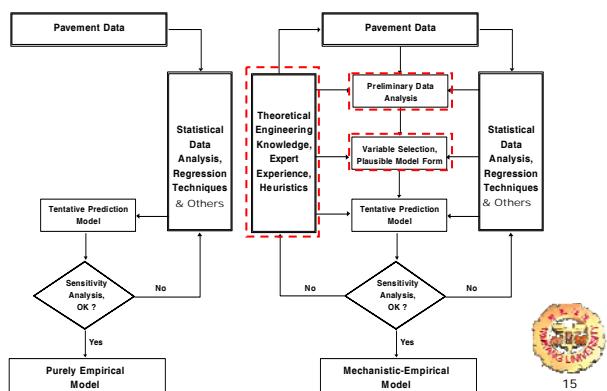
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2. Bivariate and Multivariate Analysis (Scatter Plot Matrix with lowess smoother)



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3. Model Development Using Purely Empirical or Mechanistic-Empirical Concept (Lee, 1993)



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4. Preliminary Models Using Poisson Regression

- ◆ “When events of a certain type occur over time, space, or some other index of size, it is often relevant to model the rate at which events occur.” (Agresti, 1996)
- ◆ Fatigue cracking could be treated as rate data, i.e., percent of the entire lane area.
- ◆ Agresti (1996) also suggested that using Poisson regression for rate data is an appropriate decision.
- ◆ Generalized Linear Model (GLM)



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Model Development

◆ Generalized Linear Model (GLM)

$$g(E(Y | x)) = g(\mu) = \beta_0 + \sum_{i=1}^k \beta_i x_i = \eta(x)$$

Distribution	Link Function	Variance
Normal/Gaussian	μ	1
Binomial	$\log(\mu/(1-\mu))$	$\mu(1-\mu)/n$
Poisson	$\log(\mu)$	μ
Gamma	$1/\mu$	μ^2
Inverse Normal/Gaussian	$1/\mu^2$	μ^3
Quasi	$g(\mu)$	$V(\mu)$



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Model Development

(4a) Preliminary GLM models for all zones

$$\ln(FC) = -7.455 + 0.121 * age + 0.00168 * kesal + 0.00269 * precip + 0.0473 * temp + 12319.5 * epsilon.t + 0.0133 * ft$$

Statistics : R²=0.447 , SEE=2.882 , n=176

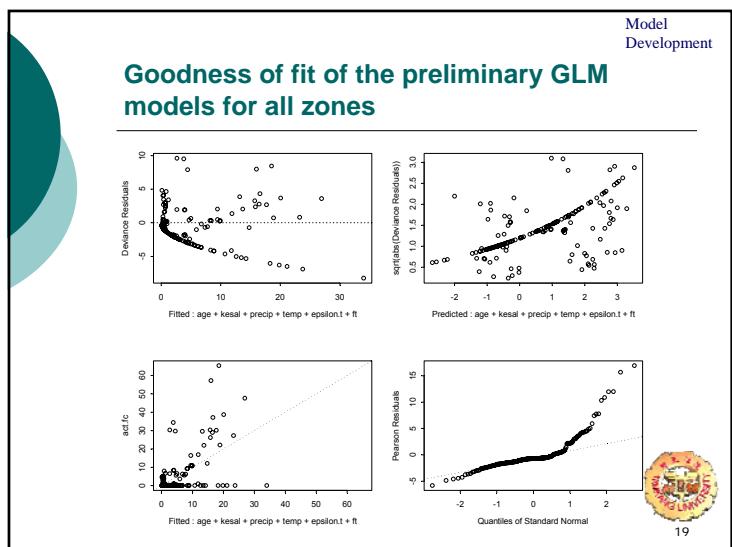
→ Convert to conventional model form:

$$FC = \exp \left[-7.455 + 0.121 * age + 0.00168 * kesal + 0.00269 * precip + 0.0473 * temp + 12319.5 * epsilon.t + 0.0133 * ft \right]$$

Statistics : R²=0.3352 , SEE=8.741 , n=176



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Model Development

(4b) Separate by different climatic zones

Variables	First Runs					Second Runs	
	wet	dry	freeze	non-freeze	dry	Freeze	
age	+	+	-*	+	+	removed	
kesal	+	-*	-*	+	removed	removed	
precip	+	+	+	+	+	+	
temp	+	+	+	+	+	removed	
epsilon.t	+	---	+	+	+	+	
ft	+	+	負*	+	+	removed	
visco						-	
temp.range						+	

---: Insignificant
 -: negative correlation
 +: positive correlation
 *: different from expectation



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Model Development

- Wet Zones (Statistics: $R^2=0.452$, SEE=3.137, n=123)

$$(FC)_{wet} = \exp \left[-6.539 + 0.078 * age + 0.00187 * kesal + 0.000673 * precip + 0.0914 * temp + 15097 * epsilon.t + 0.0272 * ft \right]$$
- Dry Zones (Statistics: $R^2=0.421$, SEE=1.117, n=53)

$$(FC)_{dry} = \exp \left[-48.411 + 0.119 * age + 0.025 * precip + 1.774 * temp + 2729 * epsilon.t + 0.0272 * ft \right]$$
- Freeze Zones (Statistics: $R^2=0.498$, SEE=1.624, n=86)

$$(FC)_{freeze} = \exp \left[-5.944 + 0.00583 * precip + 41.768 * epsilon.t - 0.002 * visco + 0.4 * temp.range \right]$$
- Non-Freeze Zones (Statistics: $R^2=0.577$, SEE=2.99, n=90)

$$(FC)_{nofreeze} = \exp \left[-7.87 + 0.102 * age + 0.00219 * kesal + 0.00102 * precip + 0.0472 * temp + 15172 * epsilon.t + 0.0476 * ft \right]$$

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Model Development

5. Improved Models Using Additional Modern Regression Techniques

- Generalized Additive Models (GAM)

$$g(E(Y | x)) = g(\mu) = \alpha + \sum_{i=1}^p f_i(x_i) = \eta(x) \quad \text{var}(Y) = \phi V(\mu)$$

- Box-Cox (1964) Power Transformation

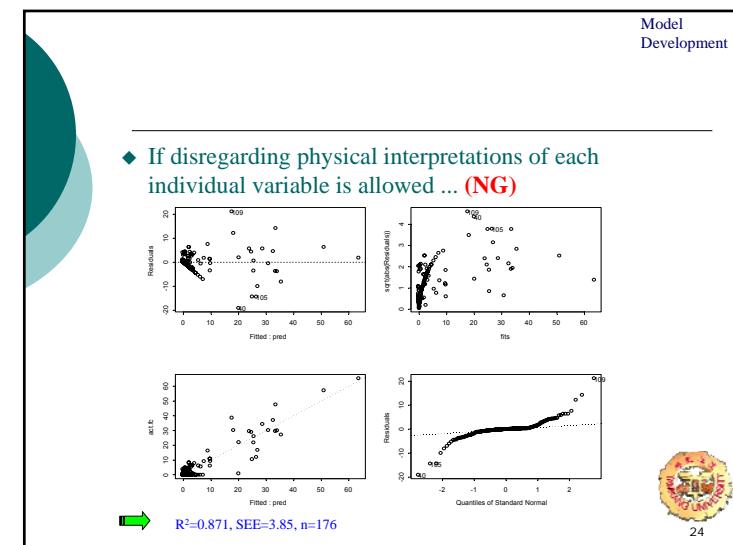
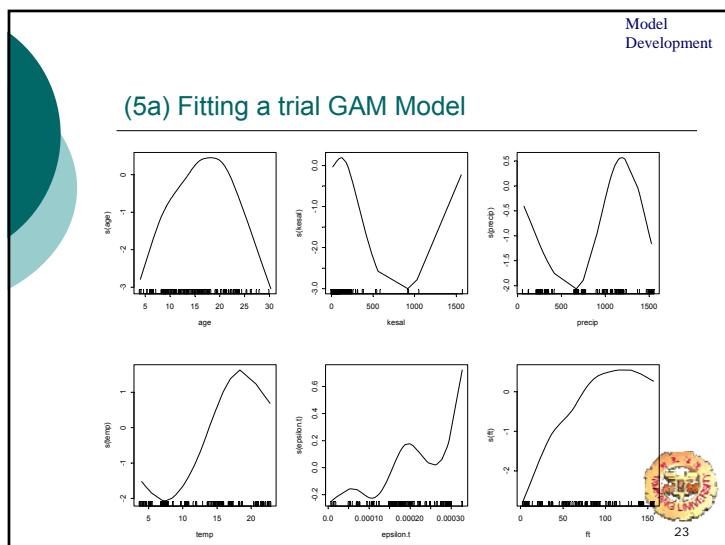
$$y_i^{(\lambda)} = \begin{cases} y_i^\lambda & \text{if } \lambda \neq 0 \\ \log(y_i) & \text{if } \lambda = 0 \end{cases}$$

$$L(\lambda) = \begin{cases} n \log (| \lambda |) - \frac{n}{2} \log (RSS_\lambda) + n (\lambda - 1) \log [GM(y)] & \text{if } \lambda \neq 0 \\ -\frac{n}{2} \log (RSS_\lambda) - n \log [GM(y)] & \text{if } \lambda = 0 \end{cases}$$

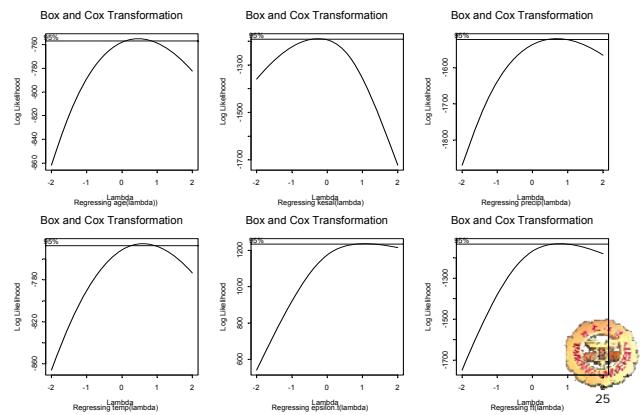
$$GM(y) = \prod_{i=1}^n (y_i)^{1/n}$$

Note: λ values from -2 to +2 are recommended

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(5b) Striving to find a monotonic power transformation function



Model
Development

(5c) Fitting a tentative GLM model

- ◆ Poisson Distribution, and quasi-likelihood estimation method, i.e., `quasi(link="log", var = "mu")`

	Value	Std. Error	t value
(Intercept)	-18.0811	3.313942	-5.45606
I(age^0.5)	0.943158	0.169463	5.565576
log10(kesal)	0.832189	0.294079	2.829816
I(precip^0.5)	0.121099	0.023639	5.122833
I(temp^0.5)	0.869489	0.387165	2.245785
I((epsilon.t * 1000)^2)	31.48981	4.30888	7.308119
log10(ft)	3.241608	1.011386	3.205116

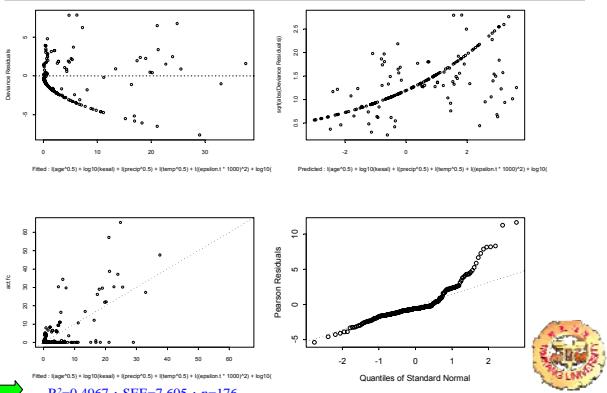
$$FC = \exp[-18.08 + 0.943 * \sqrt{age} + 0.832 * \log(kesal) + 0.121 * \sqrt{precip} \\ + 0.869 * \sqrt{temp} + 31.489 * ((\epsilon * 1000)^2) + 3.242 * \log(ft)]$$

Statistics : $R^2=0.4967$, SEE=7.605, n=176



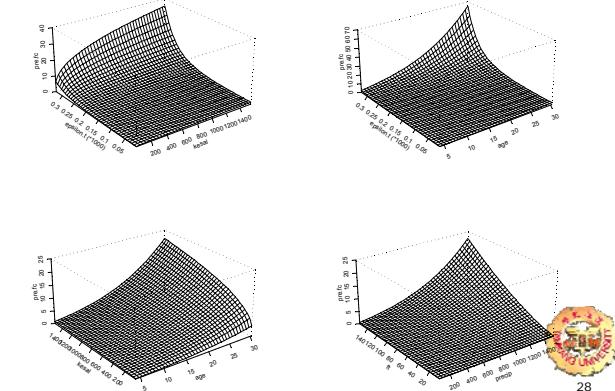
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6. Goodness of Fit of the Proposed Model



Model
Development

7. Sensitivity Analysis of the Proposed Model



Model
Development

VI. Concluding Remarks

- Existing models for fatigue cracking predictions are inadequate using LTPP Database
- Relatively skewed distribution was identified, indicating that **normality assumption is inappropriate**
- GLM and GAM along with assumption of Poisson distribution and quasi-likelihood estimation method were adopted
- By eliminating insignificant and inappropriate parameters repeatedly, the resulting model **only includes kesal, age, precip, temp, epsilon.t, and ft** for predicting fatigue cracking
- Examined the goodness of fit
- Conducted sensitivity analysis
- Further Improvements are possible and recommended



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