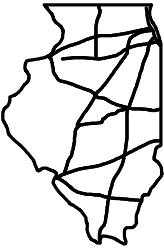



Development of Performance Prediction Models for Illinois Continuously Reinforced Concrete Pavements



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Objectives

- € Provide IPFS Data retrieval guidelines
- € Use of modern regression techniques
- € Propose a new modeling approach
- € Develop a CRCP performance prediction model
- € Study the effect of different steel placement methods

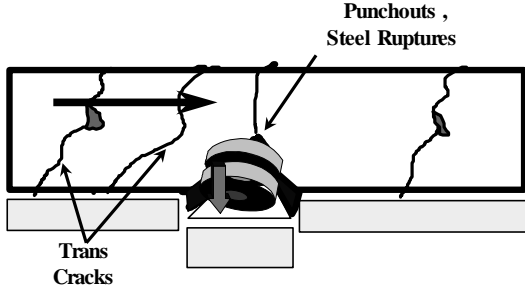


Illinois Interstate Highway system

- € Illinois Pavement Feedback System (IPFS, 1977-91)
- € 1.750 two-directional miles of multilane pavements
- € Approximately 80 % originally constructed as CRCP
- € 0.2 to 3.5 millions EASL/year,
- € wet-freeze climate, poor subgrade




Localized Failures of CRCP




Challenges Facing IDOT

- € Estimate major distress quantities for preventive design and rehab planning
- € Study the effect of different steel placement methods: Tubes € vs. Chairs




Data Preparation

- € Implementation of IPFS DB
 - Original design, Constr., Historical traffic, Distress, Condition, Maint & Rehab data
 - 1263 Interstate Sections
 - IDOT Districts, Central offices, & U of I have on-line access
- € Old Vandalia, Illinois experimental study (0.3 to 1.0% steel)
- € NOMAD2 databases ==> PC-SAS datasets




Design and Climatic Variables

- € Must consider rehab history over time
 - Distress years: 1977, 79, 82, 85, 87, 89
 - Retrieve all distress data before any ACCL
- € Steel reinforcement data
 - #5 & #6 bars before & after 1981
 - Chairs & Tubes construction before & after Sep. 1970
- € Other data
 - Drainage, base type, and environmental data




Traffic Calculation and Estimation

- € IDOT traffic maps (ADT, SU, MU)
- € Published once every 4 years
- € Automatically estimate compound traffic (ADT, EASL) growth rates
 - By performing a huge array of reg. analysis (till latest 1987 data)
 - Estimate future traffic if necessary




Distress Quantities

- € Zero or missing distresses
 - Unrecorded zero distresses or
 - Sections not actually surveyed
- € D-cracked pavements
- € Total failures
 FAIL=8.8 PATCH + PUNCH + MHPOT + NTCRK
 PATCH in %area (8 sq-ft/patch)
 others in #/mile
- € Crack spacing to check distress data




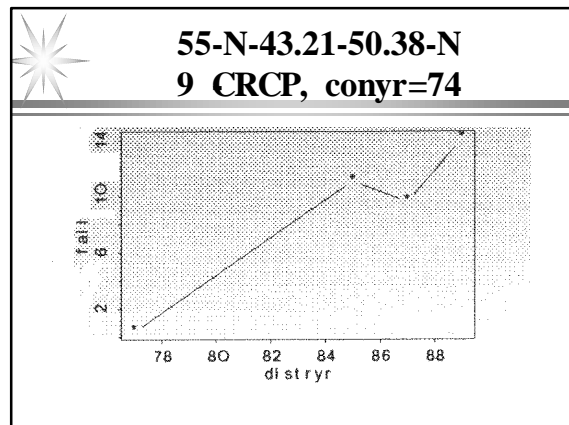
Additional Data

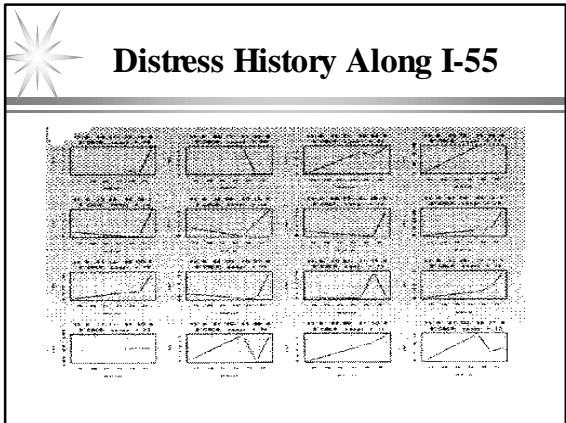
- € Old Vandalia Experimental Study
 - Constructed in 1947-48 on US 40
 - Eight sections of 7-in and 8-in. CRCP
 - 0.3, 0.5, 0.7, 1.0% steel
- € Carried 4.27 millions EASLs in 20 years
- € Many failures in 0.3 and 0.5% steel in 7-in sections




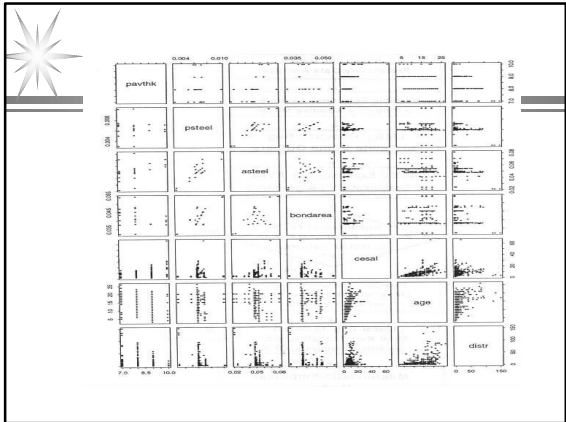
Preliminary Data Analysis

- € Distress history and additional codes



- ### Preiiminary Data Analysis (Continue ..)
- € Final database with Non-D 0 Cracked CRCP (586 data points)
 - € Correlation of Variables
 - Matrix plots of important variables: Fail, thickness, %steel, AGE, CESAL
 - Collinearity between AGE and CESAL
 - € Factor Space and Limitations
 - € Variability of the data
- 



Proposed New Predictive Modeling Approach

Traditional Regression Techniques

Linear Regression

$$y = S_0 + S_1X_1 + S_2X_2 + \dots + S_{p-1}X_{p-1} + V$$

Nonlinear Regression

$$y = F(S_0, S_1, S_2, \dots, S_{p-1}, X_1, X_2, X_{p-1}) + V$$

Minimize $RSS(S) = \sum_{i=1}^n (y_i - \hat{y}_i)^2$

Modern Regression Techniques (in S-PLUS for Windows)

LMS, Robust Regression

(Least Median Squared Reg.)
for outlier detection by Rousseeuw, 1984

$$y = S_0 + S_1x_1 + S_2x_2 + \dots + S_{p-1}x_{p-1} + V$$

Minimize $RMS(\hat{S}) = \text{med}_{M1} \sum_{i=1}^n r_i^2(\hat{S})$

ACE, Expectation Algorithm

(Alternating Conditional Expectations)
for maximizing the R-square by Breiman and Friedman, 1985

$$\hat{y}(y) = W_1(x_1) + W_2(x_2) + \dots + W_p(x_p) + V$$

Min $e^2(\hat{y}, W_1, \dots, W_p) = \frac{E[\sum_{i=1}^n (y - \hat{y}_i)^2]}{E[\sum_{i=1}^n y^2]}$

Proposed New modeling Approach

- € Conduct preliminary data analysis
- € Assume a plausible additive model (variable selection)
- € Apply Robust (LMS) Regression
- € Apply Expectation & Stabilization algorithms

AVAS, Stabilization Algorithm

(Additivity and Variance Stabilization)
to improve the fit and achieve constant variance by Tibshirani, 1987

$$\hat{y}(y) = W_1(x_1) + W_2(x_2) + \dots + W_p(x_p) + V$$

To Achieve Two Goals :

$$E[y | x_1, x_2, \dots, x_p] = \sum_{i=1}^p W_i(x_i)$$

$$VAR [y | \sum_{i=1}^p W_i(x_i)] = \text{Constant}$$

Proposed New modeling Approach

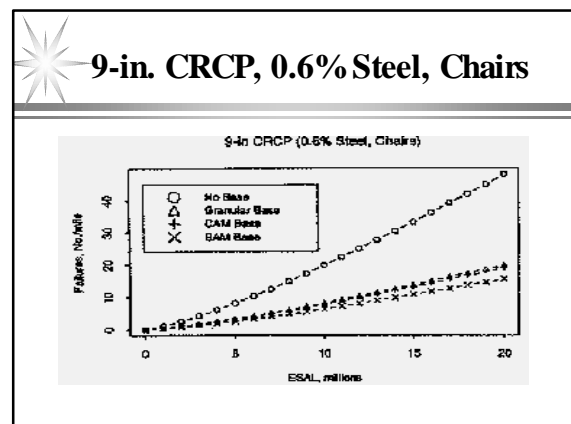
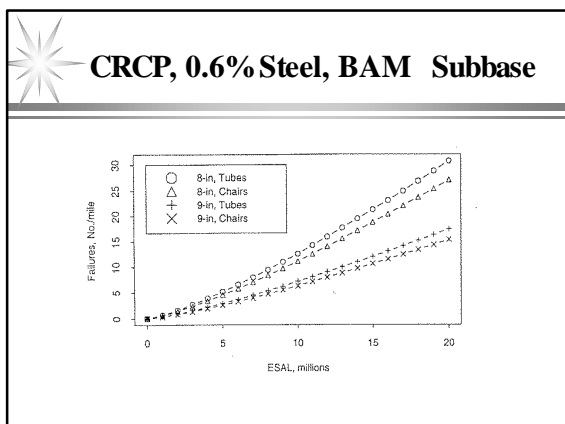
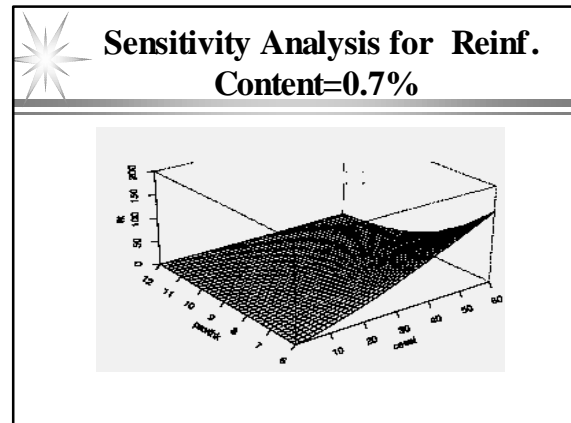
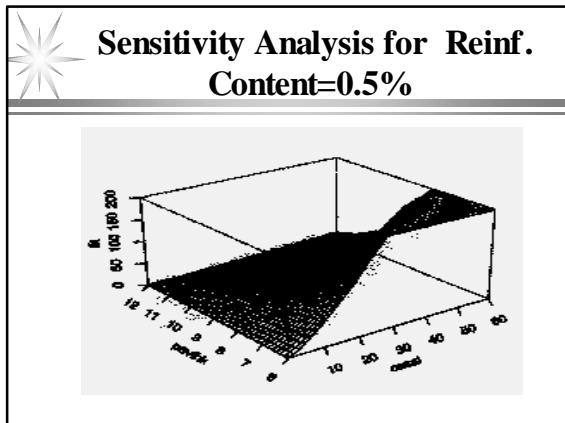
- € Check the goodness-of-the-fit, plausible functional form, detected outliers (Repeat above steps if necessary)
- € Apply traditional linear and nonlinear regression and transformation techniques
- € Obtain final regression statistics and check sensitivity analysis (Repeat if necessary)

Proposed CRCP Performance Prediction Model

$$\ln(\text{FAIR}) = 6.8004 + 0.0334 \cdot \text{PAVTHK}^2 + 6.5858 \cdot \text{PSTEEL} - 1.2875 \cdot \ln(\text{CESAL}) + 1.1408 \cdot \text{BAM} + 0.9367 \cdot \text{CAM} + 0.8909 \cdot \text{GRAN} + 0.1258 \cdot \text{CHAIRS}$$

Statistics $R^2 = 0.44$, SEE = 106, NN = 408

- € Best among several dozen models developed
- € Very high variations still present
- € Demonstrated the proposed modeling approach




Discussion

- € In the proposed model;
 - The use of chairs does result fewer total failures
 - However, the difference is not highly significant
- € Previous research performed in Illinois on I-70 has shown:
 - The depth of reinforcing steel greatly effects the crack width, breakdown of cracks, and failures
 - Tube placement method would likely have a greater chance for more failures
- € BAM base has the best overall performance
- € CAM base has about the same effect as granular base


Conclusions

- € Conducted the first attempt to use the in-service IPFS Database for research purpose
- € Provided data retrieval guidelines
 - IPFS DB & old Vandalia experimental sections used
- € Conducted a preliminary data analysis
 - For data cleaning
 - Assessing the variability of the data
 - Understanding the interrelationships among variables




Conclusions(continue...)

- € Proposed a new predictive modeling approach
 - Using several modern regression techniques
- € Developed a CRCP performance prediction
 - Failures=F(CEASL, Slab thickness, %Steel, Steel placement methods, Base Type)
- € Conducted a sensitivity analysis
- € The effect of various factors appeared to be very reasonable!



Recommendations


- € Introduce more mechanistic variables
- € Conduct a full-length survey (or at least 20-30 percent) to reduce data variability
- € Improve the proposed predictive model
- € Consider **D** Cracked Pavements




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