

1. Report No. FHWA/TX-82/26+251-1		2. Government Accession No.		3. Recipient's Catalog No.	
4. Title and Subtitle THE USE OF ROAD PROFILE STATISTICS FOR MAYSMETER CALIBRATION				5. Report Date August 1982	
				6. Performing Organization Code	
7. Author(s) David W. McKenzie, W. Ronald Hudson, and C. E. Lee				8. Performing Organization Report No. Research Report 251-1	
9. Performing Organization Name and Address Center for Transportation Research The University of Texas at Austin Austin, Texas 78712-1075				10. Work Unit No.	
				11. Contract or Grant No. Research Study 3-8-79-251	
				13. Type of Report and Period Covered Interim	
12. Sponsoring Agency Name and Address Texas State Department of Highways and Public Transportation; Transportation Planning Division P. O. Box 5051 Austin, Texas 78763				14. Sponsoring Agency Code	
15. Supplementary Notes Study conducted in cooperation with the U. S. Department of Transportation, Federal Highway Administration Research Study Title: "Deployment of a Digital Road Profilometer"					
16. Abstract The use of profile statistics from designated road sections has proved to be an effective method of Maysmeter calibration. For best results, the statistic used must be tailored so that it quantifies what Maysmeters, as a class, actually measure regardless of the intended use of the measurements. For this purpose, a family of profile summary statistics based on the concept of Root-Mean-Square Vertical Acceleration (RMSVA) was developed. Some of these statistics have been found to correlate strongly with Maysmeter response. Moreover, the relationships are linear, much as the relationships between different Maysmeter units are linear. When selected RMSVA indices are combined to form a serviceability index representing "Maysmeter roughness," an effective calibration standard results. This RMSVA-based reference, termed SIV, exhibits distinct advantages over the profile statistic (SI2) used previously for calibration. First, because of its higher correlation with Maysmeter readings, SIV allows more accurate calibrations; that is, two Maysmeters calibrated at different times are more likely to agree when measuring the same road. Second, because of its relative simplicity, SIV is a more consistent and precise measurement of roughness when considered as a profilometer measurement. The repeatability of measurements, as shown by pairs of successive profilometer runs and also by consistency on sections that change very slowly, is much improved. Since it is less sensitive to the characteristics of profilometer hardware and is simple to compute, RMSVA is well suited for transition to the new digital profilometer which is being purchased by the Texas State Department of Highways and Public Transportation (SDHPT).					
17. Key Words road profile, statistics, Maysmeter, calibration, root-mean-square vertical acceleration (RMSVA), serviceability index, roughness			18. Distribution Statement No restrictions. This document is available to the public through the National Technical Information Service, Springfield, Virginia 22161.		
19. Security Classif. (of this report) Unclassified		20. Security Classif. (of this page) Unclassified		21. No. of Pages 154	22. Price

THE USE OF ROAD PROFILE STATISTICS FOR
MAYSMETER CALIBRATION

by

David W. McKenzie
W. Ronald Hudson
C. E. Lee

Research Report Number 251-1

Deployment of a Digital Road Profilometer
Research Project 3- 8-79-251

conducted for

Texas
State Department of Highways and Public Transportation

in cooperation with the
U. S. Department of Transportation
Federal Highway Administration

by the

CENTER FOR TRANSPORTATION RESEARCH
BUREAU OF ENGINEERING RESEARCH
THE UNIVERSITY OF TEXAS AT AUSTIN

August 1982

The contents of this report reflect the views of the authors, who are responsible for the facts and the accuracy of the data presented herein. The contents do not necessarily reflect the official views or policies of the Federal Highway Administration. This report does not constitute a standard, specification, or regulation.

PREFACE

This is the first report presenting results from Research Project 3- 8-79-251, entitled "Deployment of a Digital Road Profilometer." The study has two primary objectives: (1) to provide continuing support to the Texas State Department of Highways and Public Transportation (SDHPT) in all projects requiring profilometer data and (2) to assist SDHPT in the purchase and adaptation of a new digital profilometer system while insuring that Maysmeter calibration and other projects are not adversely affected by the transition to the new hardware.

The Surface Dynamics Profilometer (SDP) was originally developed under Research Studies 3-8-63-73, entitled "Development of a System for High-Speed Measurement of Pavement Roughness," and 3-8-71-156, entitled "Surface Dynamics Road Profilometer Applications."

A significant result of these studies was the development of a set of equations which can be used to predict a serviceability index that is based on road profile statistics. Present Serviceability Ratings (PSR) had been obtained from a 1968 subjective rating experiment on Texas roads; then the profile-based Serviceability Indices (SI) made it possible to interpret profilometer data in quantitative terms that relate more directly to ride quality or roughness. One of those indices, SI₂, has served for a number of years as a relatively stable reference statistic for calibrating the SDHPT Maysmeters.

This report describes work by The University of Texas at Austin's Center for Transportation Research (CTR) to upgrade the existing Maysmeter calibration program by developing an improved profile statistic for use as a calibration standard. SI2 was not an optimal reference, primarily because of the presence of long wavelength roughness components that are not measurable by the Maysmeter. The new statistic, which results from a Maysmeter simulation that is based on Root-Mean-Square Vertical Acceleration (RMSVA), allows more accurate calibrations and will be more easily adapted to on-board computation with the new digital profilometer that was delivered in 1982.

The authors are pleased to acknowledge the work of Maitree Srinarawat, who assisted in much of the original data analyses through which the relevant RMSVA indices were derived. Also appreciated is the kind assistance of the State Department of Highways and Public Transportation representative, Brad Hubbard (D-10 Research), who furnished the Maysmeter data crucial to this study and who was first to point out certain deficiencies in the original calibration program.

Finally, special thanks are due to the CTR staff members who participated in this project, especially to Prentiss Riddle for his programming efforts and Elaine Hamilton, Barbara Allen, and Lyn Gabbert for their typing and proofing of the manuscript.

David W. McKenzie

W. Ronald Hudson

Clyde E. Lee

LIST OF REPORTS

Report No. 251-1, "The Use of Road Profile Statistics for Maysmeter Calibration," by David W. McKenzie, W. Ronald Hudson, and Clyde E. Lee, discusses the development of a set of summary statistics, termed Root-Mean-Square Vertical Acceleration (RMSVA) indices, useful for evaluating pavement roughness and predicting Maysmeter response. The SDHPT Maysmeter calibration program is revised to incorporate a linear calibration model based on an RMSVA-derived Maysmeter simulation.

ABSTRACT

The use of profile statistics from designated road sections has proved to be an effective method of Maysmeter calibration. For best results, the statistic used must be tailored so that it quantifies what Maysmeters, as a class, actually measure regardless of the intended use of the measurements.

For this purpose, a family of profile summary statistics based on the concept of Root-Mean-Square Vertical Acceleration (RMSVA) was developed. Some of these statistics have been found to correlate strongly with Maysmeter response. Moreover, the relationships are linear, much as the relationships between different Maysmeter units are linear. When selected RMSVA indices are combined to form a serviceability index representing "Maysmeter roughness," an effective calibration standard results. This RMSVA-based reference, termed SIV, exhibits distinct advantages over the profile statistic (SI2) used previously for calibration.

First, because of its higher correlation with Maysmeter readings, SIV allows more accurate calibrations; that is, two Maysmeters calibrated at different times are more likely to agree when measuring the same road. Second, because of its relative simplicity, SIV is a more consistent and precise measurement of roughness when considered as a profilometer measurement. The repeatability of measurements, as shown by pairs of successive profilometer runs and also by consistency on sections that change very slowly, is much improved. Since it is less sensitive to the

characteristics of profilometer hardware and is simple to compute, RMSVA is well suited for transition to the new digital profilometer which is being purchased by the Texas State Department of Highways and Public Transportation (SDHPT).

SUMMARY

One of the more important applications of the Surface Dynamics Profilometer (SDP) is to provide a stable calibration reference for response-type roughness measuring instruments. The latter devices, of which the Mays Ride Meter (Maysmeter) is typical, are relatively inexpensive and are used by many agencies for routine pavement monitoring. In Texas, a Maysmeter is calibrated by running the unit on approximately 30 designated flexible pavement road sections near Austin and then using statistical means to correlate its measurements with appropriate data obtained on those same sections with the profilometer.

The use of the profilometer as a reference vehicle has avoided the difficulty of trying to maintain a "stable" Maysmeter for this purpose; however, the problem of relating profile data to Maysmeter measurements must be solved if accurate calibrations are to result.

Two approaches are possible for developing profile statistics suitable for both roughness measuring device calibration and general roughness evaluation. The first is dynamic modelling of a hypothetical device with certain physical constants pre-defined and with sequences of profile evaluations taken as system input. For example, the Quarter Car Index (QI), which has been used successfully for Maysmeter calibration in Brazil, was so developed. Such indices are useful, however, not as simulations of particular instruments but simply as profile summary statistics whose

required high correlation with the target class of devices must be shown experimentally.

The alternative statistical approach which is described in this report is to obtain data from response-type roughness measuring instruments on representative road sections and then use regression techniques to select a profile statistic which the instruments are capable of measuring reliably.

In early 1978 a profile statistic which in principal relates to root-mean-square vertical acceleration (RMSVA) was investigated for use as a road roughness measure. It was able to describe quite well the behavior of eight Maysmeters (mounted on five trailers and three cars) all run on the Austin test sections in late 1977. A Maysmeter simulation statistic was thereby derived. The corresponding serviceability index, SIV, has since proved to be a definite improvement over the calibration standard, SI2, that was used previously. The latter statistic is based in part on certain power spectral estimates to which the Maysmeters are not sensitive.

For a period of two years, both SIV and SI2 were recorded during the quarterly profilometer runs of the test sections. Although the overall mean and variability of the sections were about the same for both indices, the SIV differences between repeated profilometer runs were only one-fourth to one-third as great as the SI2 differences. Furthermore, SIV is a more reliable measurement of roughness, as shown by consistency on sections having no major maintenance and experiencing little or no overall change in roughness during the period of study. The serviceability of 19 essentially unaltered sections declined an average of only 0.2 unit per year.

Although SIV was finally adopted as the calibration standard during January 1980, it was adopted in a provisional way that did not require modification of MRMCAL, the program that processes the calibration session data and produces a chart relating raw profilometer readings to serviceability estimates. The original version of MRMCAL employs a nonlinear regression algorithm. The recommended method is to use a revised program, MRMCAL2, to perform a simple linear fitting of the Maysmeter readings to an RMSVA-derived Maysmeter simulation.

Although the Maysmeter calibration problem motivated the development of RMSVA roughness indices, careful monitoring of the Austin test sections and other pavements has revealed surface properties that could never be detected by Maysmeters or by devices which reduce roughness evaluations to a single number. The RMSVA indices computed from a road profile can provide a "signature" that reflects roughness over a broad range of profile wavelengths. Distinctive signatures corresponding to certain pavement classes or types of deterioration have been tentatively identified and their interpretation remains as a promising subject for future research.

IMPLEMENTATION STATEMENT

It is recommended that the current SDHPT Maysmeter calibration program be fully revised to employ the linear calibration model described herein. A new version of the computer program which produces Maysmeter calibration tables is presented for replacement of the version now in use. The recommended change will simplify existing procedures and will result in more accurate calibrations.

METRIC CONVERSION FACTORS

Approximate Conversions to Metric Measures

Symbol	When You Know	Multiply by	To Find	Symbol
LENGTH				
in	inches	2.5	centimeters	cm
ft	feet	30	centimeters	cm
yd	yards	0.9	meters	m
mi	miles	1.6	kilometers	km
AREA				
in ²	square inches	6.5	square centimeters	cm ²
ft ²	square feet	0.09	square meters	m ²
yd ²	square yards	0.8	square meters	m ²
mi ²	square miles	2.6	square kilometers	km ²
	acres	0.4	hectares	ha
MASS (weight)				
oz	ounces	28	grams	g
lb	pounds	0.45	kilograms	kg
	short tons (2000 lb)	0.9	tonnes	t
VOLUME				
tsp	teaspoons	5	milliliters	ml
Tbsp	tablespoons	15	milliliters	ml
fl oz	fluid ounces	30	milliliters	ml
c	cups	0.24	liters	l
pt	pints	0.47	liters	l
qt	quarts	0.95	liters	l
gal	gallons	3.8	liters	l
ft ³	cubic feet	0.03	cubic meters	m ³
yd ³	cubic yards	0.76	cubic meters	m ³
TEMPERATURE (exact)				
°F	Fahrenheit temperature	5/9 (after subtracting 32)	Celsius temperature	°C

Approximate Conversions from Metric Measures

Symbol	When You Know	Multiply by	To Find	Symbol
LENGTH				
mm	millimeters	0.04	inches	in
cm	centimeters	0.4	inches	in
m	meters	3.3	feet	ft
m	meters	1.1	yards	yd
km	kilometers	0.6	miles	mi
AREA				
cm ²	square centimeters	0.16	square inches	in ²
m ²	square meters	1.2	square yards	yd ²
km ²	square kilometers	0.4	square miles	mi ²
ha	hectares (10,000 m ²)	2.5	acres	
MASS (weight)				
g	grams	0.035	ounces	oz
kg	kilograms	2.2	pounds	lb
t	tonnes (1000 kg)	1.1	short tons	
VOLUME				
ml	milliliters	0.03	fluid ounces	fl oz
l	liters	2.1	pints	pt
l	liters	1.06	quarts	qt
l	liters	0.26	gallons	gal
m ³	cubic meters	35	cubic feet	ft ³
m ³	cubic meters	1.3	cubic yards	yd ³
TEMPERATURE (exact)				
°C	Celsius temperature	9/5 (then add 32)	Fahrenheit temperature	°F

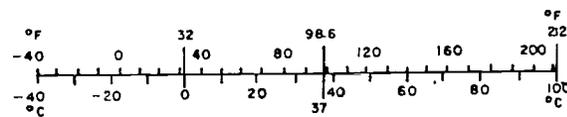
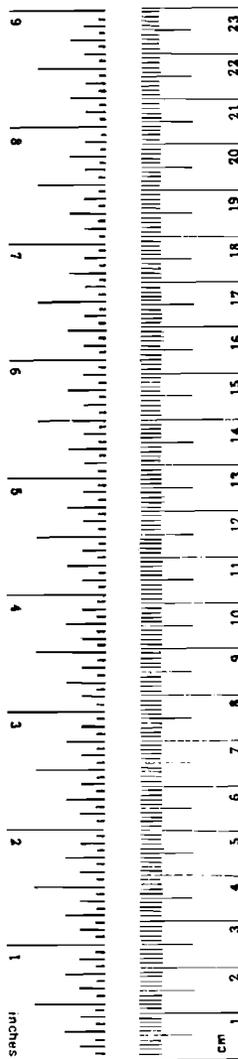


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CHAPTER 1. BACKGROUND AND APPLICATIONS OF THE SURFACE DYNAMICS PROFILOMETER

The Texas State Department of Highways and Public Transportation (SDHPT) has long had a vital interest in assuring the quality of its road system and in predicting pavement performance. To this end, the Surface Dynamics Profilometer (SDP) was introduced into the Department in early 1967. This system, which includes road profile measuring equipment along with data analysis programs that have been developed at The University of Texas Center for Transportation Research (CTR), continues to be an essential part of the state-wide program for planning, designing, operating, and maintaining the highway network.

As a roughness measuring device, the SDP is unique in that numerous roughness and performance statistics can be extracted from the profile data. The information can include ride quality or serviceability, evidence of various kinds of pavement distress, and even simulations of response-type roughness measuring devices. Accurate road profile measurements can be made at high speed, and because the measurements are essentially independent of vehicle suspension, they have shown to be very stable in time. For this reason, the SDP has functioned for the past eight years as the standard reference device for calibrating the SDHPT Maysmeters.

The original profilometer hardware, which is mounted in a 1966 General Motors passenger vehicle, records all profile data in analog form.

Consequently, a laboratory-based analog-to-digital process is required before profile summary statistics can be computed. This equipment, while still operational, is soon to be augmented and eventually replaced with a much improved digital system, which will provide usable roughness data while still in the field and will produce digital tapes for analysis with existing computer software.

SUMMARY OF PAST RESEARCH

The first SDP purchased by the SDHPT was manufactured by K. J. Law Engineering Company under patents of General Motors Corporation. The device was purchased in early 1967 after a thorough investigation of existing road roughness measuring equipment, as described in Research Report No. 73-1, "High-Speed Road Profile Equipment Evaluation." Subsequent to a shakedown phase in which various operating problems were found and eliminated, a road profile measuring system was developed which consisted of both the SDP and a companion A-D computer facility for converting analog profile data to digital form for computer analysis. The details of this system are described in Research Report No. 73-2, "A Profile Measuring, Recording, and Processing System." Research Report No. 73-4, "An Analog-to-Digital System," describes the Hewlett-Packard 2115 A-D system that has been operated by the Texas State Department of Highways and Public Transportation since 1969.

In order to obtain a meaningful roughness description from a measured road profile, efforts were made to correlate various summary statistics, such as slope variance, with Present Serviceability Rating (PSR). PSR is defined as the average subjective rating, on a scale from zero to five, of a road's

ability to serve high-speed, high-volume, mixed traffic in its present condition. In the summer of 1968, two large-scale rating sessions were conducted by Center for Highway Research. Representative pavement sections were rated by a human panel after the profiles had been measured. A discussion of the rating methods and the first PSR prediction equations which resulted is provided in Research Report No. 73-3, "Pavement Serviceability Equations Using the Surface Dynamics Profilometer."

Under Research Study 3-8-71-156, improved analytical methods for obtaining more realistic and useful descriptions of the road profile were developed and much of the computer software which is currently in use was written. Improved PSR prediction equations were obtained, and significant progress was made toward identifying the profile characteristics which relate most directly to different types of roughness. In addition, use of the profilometer was extended to a wider variety of research and implementation areas.

In the early 1970's, the Department established a set of priority needs for roughness measurements, including the following:

- (1) assessment of pavement performance for maintenance purposes,
- (2) evaluation of the roughness of special pavement sections such as bridges and railroad crossings, and
- (3) evaluation and priority determinations of pavement rehabilitation methods, both rigid and flexible.

To help accomplish these goals, the SDHPT acquired a group of Maysmeters to function as roughness measuring devices in various locations around the

State. Maysmeters respond to road roughness in a way that is dependent on vehicle suspension and other factors that may vary over time and between devices. Research Report 156-1, "A Correlation Study of the Mays Road Meter with the Surface Dynamics Profilometer" (February 1973), documents the development of a statistical model for predicting serviceability in terms of the Maysmeter roughness measure (inches of axle displacement per mile) for a given vehicle. More important, the profilometer-derived Serviceability Index (SI) was proposed as a reasonably stable standard against which different Maysmeters could be compared. Although the calibration standard SI was eventually replaced, the basic calibration procedures introduced at that time are still in use at SDHPT.

In Research Report 156-2, "The Use of Spectral Estimates for Pavement Characterization" (August 1973), a method for predicting the serviceability in terms of root-mean-square profile wavelength amplitudes is presented. The multiple regression model (SI2) contains 22 terms involving power spectral estimates extracted from the road profile. The model was considered successful in that 89 percent of the road-to-road variation in rating panel serviceability data was explained in terms of roughness amplitudes in the wavelength range of 8 to 87 feet. The success of the SI2 regression model encouraged a deeper investigation into the relationship between roughness and serviceability. In particular, this statistic alone was still not adequate for answering such pertinent questions as the following:

- (1) What is the within-section roughness variation? For example, a given RMS roughness amplitude could be due to either a single large bump or many waves of smaller amplitude, etc.

- (2) What is the effect of vehicle roll (due to right-left wheelpath deviation difference) on ride quality?
- (3) What relationships can be established between specific kinds of roughness and known types of pavement distress?
- (4) What features of the road roughness do users find most objectionable? What are their individual correlations with serviceability?

The research that followed helped to resolve some of these questions. This work is described in Research Reports 156-3, "Analysis of Characteristic Roughness Patterns in Pavements and the Relationship Between Roughness and Pavement Distress;" 156-4, "The Characterizations of Road Roughness on Bridge Decks and the Adjoining Pavement," and finally, 156-5F, "A Study of the Relationship Between Various Classes of Road Surface Roughness and Human Ratings of Riding Quality."

The computer program ROKYRD, which analyzes and plots filtered profiles, was an important product of the research effort. Digital filtering techniques allow specific profile wavelengths to be isolated and compared. Furthermore, frequency distributions in the form of percentile roughness amplitudes for each desired waveband are provided, as are the individual and combined serviceability estimates.

In early 1978, CTR again proceeded to investigate various profile summary statistics, this time in an effort to devise a better Maysmeter calibration standard. One statistic in particular, Root-Mean-Square Vertical Acceleration (RMSVA), was able to describe especially well the behavior of

the Department's Maysmeters on the calibration test sections. The new calibration standard resulting from this work has since proved to be a definite improvement over the serviceability index SI2, which is partially based on power spectral quantities to which the Maysmeter is not sensitive.

The development of the RMSVA roughness statistics and their correlation with Maysmeters are described in this report, the first produced under Research Study 3- 8-79-251, "Deployment of a Digital Road Profilometer."

TYPICAL APPLICATIONS OF THE PROFILOMETER

A brief summary of current or recent SDHPT activities which are typical of possible future projects involving the SD Profilometer is given below:

- (1) Maysmeter Calibration - Since early 1973, the SD Profilometer has provided, on a regular quarterly basis, the serviceability estimates for approximately 30 calibration test sections. A particular Maysmeter is calibrated by running it on the test sections, then scaling its measurements to fit the most current profilometer data. (In this report, the development of an improved calibration standard based on profile statistics is described.)
- (2) Evaluation of Construction Methods - By comparing bridge surfaces constructed with longitudinal and transverse screening (Study 3-213), it was demonstrated that the latter process introduced more short-wave roughness. In a separate but similar study, the effect on roughness amplitudes of ski and tight-string procedures in hot mix laying was investigated.
- (3) Roughness at the Bridge-Pavement Interface - Both the magnitude and the frequency of occurrence of this problem have been examined by

the profilometer. The goal is to establish a probable cause of this roughness and to make recommendations for design and maintenance.

- (4) Dynamic Loading of Bridges - A computer program DYMOL uses profilometer data to predict the dynamic loading caused by a vehicle. This information has been used to suggest suitable speed limits on bridges which have experienced sagging.
- (5) Evaluation of Overlay Methods - To study the performance of asphaltic overlays on CRCP, profiles of test sections in selected counties are being monitored on a yearly basis (Project 249).
- (6) Evaluation of Rehabilitation Methods - The effect of pressure grouting on CRC pavements known to contain voids was investigated by examining profiles of test and control sections. The statistical summaries provided by ROKYRD indicated that roughness in the 25 to 50-foot wavelength range increased where grouting occurred. This supported other evidence that the slab was occasionally raised by the process.
- (7) Implementation of Rehabilitation Methods - The profilometer strip chart has been used to estimate the quantity of material needed for an efficient level-up. Software available for the new digital profilometer will simplify this procedure.
- (8) Pavement Materials - Profilometer measurements have been analyzed to determine the effect of sulfur-asphaltic mixtures on pavement quality.

- (9) Pavement Distress Mechanisms - On a semiannual basis profilometer runs were made in connection with Project 224, "Detrimental Volume Changes in Expansive Clays in Highway Subgrades."

ADVANCED FEATURES OF THE DIGITAL PROFILOMETER

In addition to providing continuing support of SDHPT projects requiring road profile data, a major objective of Research Study 3- 8-79-251 has been the purchase and adaptation of a new digital profilometer to replace the existing hardware, which has been in use for more than 10 years. The principal operational advantages of the digital model are given below.

- (1) Almost limitless range is provided for digital integration of electrical signals. Operating the old profilometer on grades sometimes overloads the integrators, thereby creating an artificial step in the profile.
- (2) Variable digital filtering is incorporated in the new equipment. The profile cutoff frequency set by the operator remains fixed as the filtering system adjusts automatically to the speed of the vehicle.
- (3) Calculation of various roughness statistics (e.g., Quarter-Car Index, RMSVA, etc.) is accomplished on board the vehicle. The main processor can be re-programmed, and new programs can be added.
- (4) An electric typewriter input-output console for printing these statistics and for recording other run information will be

provided. Also, strip chart output will be available via a digital-to-analog conversion system.

- (5) A larger and more functional van houses the equipment.
- (6) A distance measuring system using the left front wheel of the profilometer vehicle provides a more accurate speed and distance input to the digital recording system.
- (7) Self-calibration and self-checking of system operation are provided by programs in the on-board computer.
- (8) A magnetic tape format which represents each profile data point by a 24-bit binary number is utilized. This allows for a resolution of .001 inch and a range of ± 8388.608 inches.

The new system, with its improved hardware and more compact configuration, is expected to be of greater benefit to the field engineer. Thus, in addition to serving as a standard calibration device for other roughness measuring instruments, the profilometer will be used even more extensively as a direct aid for pavement design and evaluation.

CHAPTER 2. ROOT-MEAN-SQUARE VERTICAL ACCELERATION AS A MEANS FOR EVALUATING ROAD ROUGHNESS

THE MAYS METER CALIBRATION PROBLEM

Since early 1973, the Texas State Department of Highways and Public Transportation has calibrated its Maysmeters by running each unit periodically on approximately 30 pavement test sections near Austin. Each calibration session determines a relationship between a particular unit's measurements and the serviceability estimates of a Surface Dynamics Profilometer. The use of the profilometer as a reference vehicle has avoided the difficulty in trying to maintain for that purpose a typical Maysmeter whose response characteristics, it is hoped, never change. Instead, one is faced with the initial problem of relating Maysmeters to the profilometer. For example, one would expect that two Maysmeter trailers, being similarly constructed, would respond to more or less the same kinds of pavement properties. There is a good chance that their separate measurements are related by a simple, perhaps linear, function. The profilometer, on the other hand, measures only the road profile but does so with considerable accuracy and consistency. The problem is that the function relating the profile to a Maysmeter's measurements may not be simple and there is a possibility of choosing an inferior, possibly invalid, profile summary statistic as the calibration standard.

For a number of years, the approach in Texas was to calibrate each Maysmeter independently by fitting its measurements on the test sections to the equation

$$\ln M \approx \beta \left[\ln \left(\frac{5}{SI} \right) \right]^{1/\alpha} \quad (2.1)$$

where

M = is the Maysmeter measurement (inches per mile),

SI = is the current serviceability index for the section, and

β and α = are nonlinear regression coefficients.

SI is itself a prediction equation developed independently to explain the results of a 1968 subjective panel rating of 86 sections in the Dallas-Fort Worth area (Ref 1). It contains 22 terms involving power spectral estimates computed from the digitized road profile.

The form of Eq 2.1 was obtained empirically after considerable experimentation (Ref 2), and served to provide each calibrated unit with an equation,

$$\widehat{SI} = 5e^{-\left(\frac{\ln M}{\beta}\right)^\alpha} \quad (2.2)$$

to convert its readings to serviceability indices which range from 0 to 5. Ideally, the adjustment of parameters α and β through calibration would account for physical differences between units or for changes in a given vehicle's suspension, tires, etc., due to wear or replacement of parts.

Eventually, a weakness in the Texas method became apparent when it was noticed that several of the test sections were persistent regression outliers; that is, they deviated significantly from the calibration regression curves and did so with each Maysmeter that was calibrated. Also, the effect of the calibration process was somewhat unpredictable. In the period preceding a calibration session a Maysmeter would provide \widehat{SI} values of, say, 2.6, 2.7, 2.6, etc., for a succession of runs on a particular road. Afterwards, the same unit would return with new calibration parameters and provide, for the same road, \widehat{SI} values of 3.2, 3.3, 3.2, etc., when it was doubtful that either the road or Maysmeter had physically changed. More likely, the group of rougher calibration test sections, for which there had been consistently poor agreement between the profilometer and Maysmeters, had changed since the previous calibration of that unit. Initially, there was an inclination to discard some of these sections due to their disproportionate effect on the calibration parameters; however, it was discovered that plots of \widehat{SI} versus SI for different Maysmeters were almost identical with respect to the pattern of outliers. It then became clear that the calibration standard SI, not the choice of sections, was at fault as the statistic was obviously insensitive to roughness components measured with good repeatability by the Maysmeters.

Figure 1.1 is presented as a conceptual illustration of the advantage of calibrating four Maysmeters, for example, against a profile statistic which is specifically tailored to represent "MRM roughness." The responses of calibrated and uncalibrated devices are shown for four road sections with similar MRM roughness but with different present serviceability ratings (PSR). If PSR had instead been chosen as the calibration standard and the four units were "calibrated" at one time on sections 1 and 2 and at another

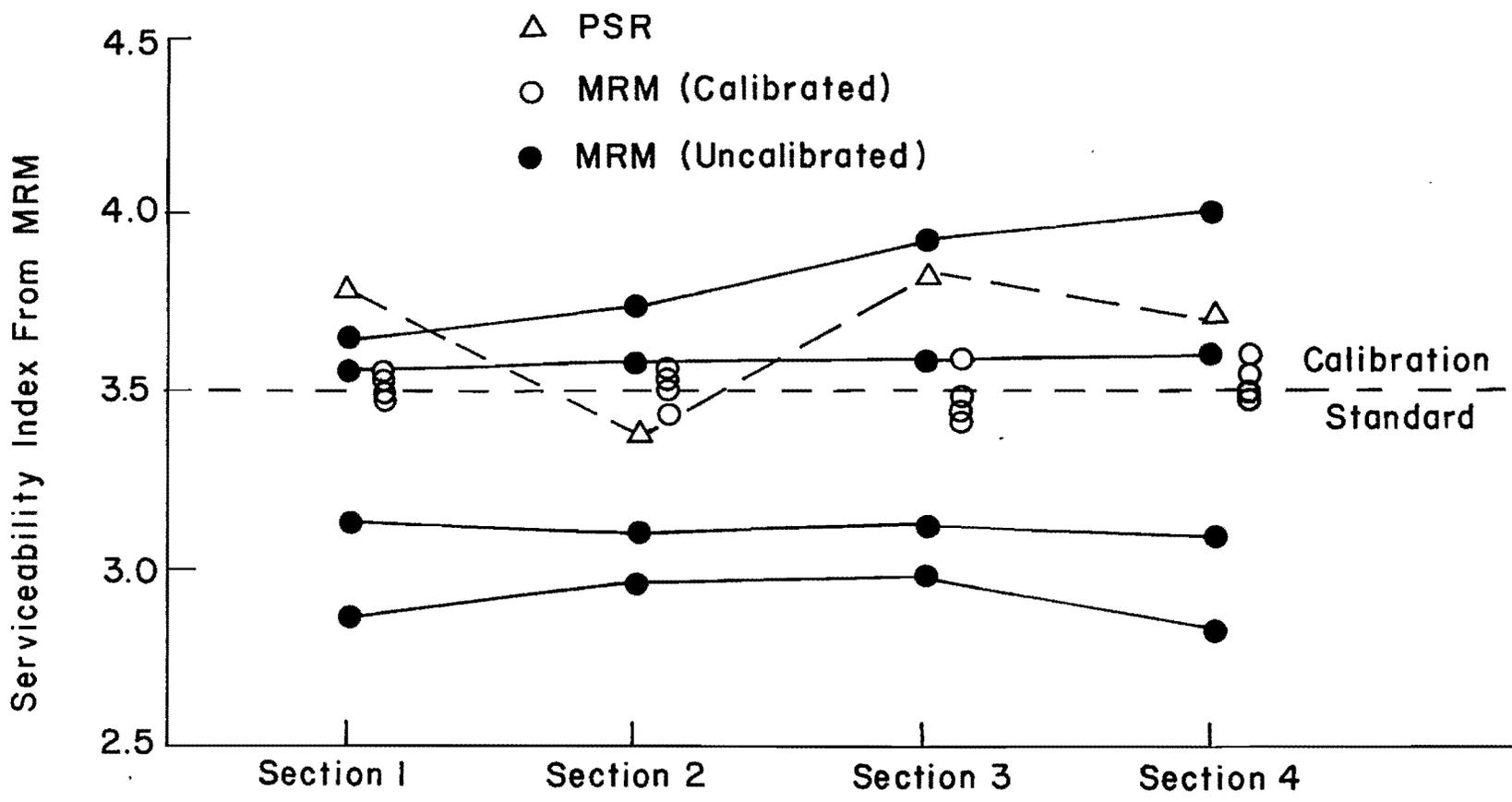


Fig 2.1. Example behavior of four Maysmeters on road sections with similar (MRM) roughness but with different present serviceability ratings (TSR).

time on sections 3 and 4, then it is clear that the units would acquire, in each calibration, a different formula for adjusting their raw readings, even if no physical changes in the devices have occurred. (Of course, an appropriate collection of road sections would have different levels of MRM roughness.) Although the procedure might succeed in bringing different units that happen to be calibrated at the same time into agreement, any variables in the assumed standard which are not measurable by the Maysmeter can only weaken the correlation between the devices and their common reference, thereby reducing the effectiveness of the calibration process.

On the other hand, if the formula resulting from the calibration were basically just a rescaling of measurements to insure the compatibility and time stability of different Maysmeters, then the essential information would not be degraded; it would always be available for deriving better relationships, for example, between MRM roughness and PSR or vehicle operating costs. While these latter correlations may be of utmost importance, they should not be confused with the prerequisite process of calibration.

The primary advantage of using a calibration reference specific for the particular kind of target device is that it results in more accurate calibrations. This means that it will be safer to assume that road A is actually rougher than road B even though the measurements indicating this were obtained at different times or with different "calibrated units." This is simply a benefit of the higher correlation between the instrument and an established reference. Pavement engineers would appreciate the ability to make valid comparisons of different pavements, which may vary only slightly with respect to apparent PSR but which can, nonetheless, reveal the effects of different maintenance practices or show significant trends over a period

of time. Many response-type measuring devices, including Maysmeters, can measure with good repeatability certain traits of a pavement surface, or profile, which would not be as reliably observed by a rating panel.

The limitations of the original SI equation are not surprising for it was not developed in view of the Maysmeter's capabilities. Instead, power spectral and cross-spectral estimates from 64 frequency bands were considered in the regression model, the goal being to find a best predictor of Present Serviceability Rating (PSR) utilizing a profilometer's measurements. The SI equation has terms for roughness amplitudes for wavelengths up to 83 feet whereas it is known that Maysmeters run at 50 mph respond to a much smaller wavelength range--say 4 to 40 feet.

Consequently, an effort was made to improve the calibration procedure by selecting an alternative profile statistic for use as a standard. Our approach, which is described in the remainder of this report, was largely empirical in that we were not especially concerned with identifying or modelling the causal chain leading from the physical road profile to the roughness readings of a particular device. The need was to isolate a feature of the profile, something conceptually simple and easily computed if possible, which tends to quantify what Maysmeters as a class do, in fact, measure.

RMSVA - INTRODUCTION AND DEFINITIONS

Improving the Texas MRM calibration method required our searching for a profile summary statistic that simulated the response of a typical Maysmeter. Previous attempts to derive a predictor of Present Serviceability Rating

(PSR), which is a subjective assessment of roughness by users, deal principally with two basic methods for deriving roughness-related quantities from a sequence of wheelpath elevations. The first is computation of the power spectrum to obtain a set of indices that depend on the frequency of irregularities in the profile. The development of the previously mentioned SI equation, in which stepwise regression is used to select terms based on spectral estimates, is described in Ref 2. The second approach is that of digital filtering in which the disturbances in a set of contiguous pass-bands are isolated and examined for their effects on PSR (Ref 4). In this case, moving root-mean-square amplitudes are computed for each band, along with 50, 75, 90, 95, and 99 percentile values, which help distinguish between roads that have roughness either spread out or highly localized.

In our search for a statistic representing Maysmeter roughness, it was found that these techniques for detailed characterization of the profile were computationally wasteful and also rather sensitive to the profilometer speed and hardware configuration at the time of measurement. Instead, a very simple statistic, which is not so critically dependent on profile measuring technique and resolution, met the basic requirements of a calibration standard: (1) that it accurately represent, or simulate, the class of devices to be calibrated and (2) that an independent, time-stable method of measuring it be available. The component profile measure can be simply described as the root-mean-square difference between adjacent profile slopes, where each slope is the ratio of elevation change to distance over a fixed distance increment. Since the required computation is equivalent to estimating the second derivative of height with respect to time for an object in contact with the profile moving horizontally at a fixed speed, this

quantity has been termed Root-Mean-Square Vertical Acceleration (RMSVA). It is defined more exactly as follows.

Let Y_1, Y_2, \dots, Y_N represent elevations of equally spaced points along one wheelpath of the profile. If s is the horizontal distance between adjacent points (sampling interval), then a simple estimate of the second derivative of Y at point i with respect to distance is

$$\begin{aligned} (S_b)_i &= \frac{(Y_{i+k} - Y_i)/ks - (Y_i - Y_{i-k})/ks}{ks} \\ &= (Y_{i+k} - 2Y_i + Y_{i-k}) / (ks)^2, \end{aligned} \quad (2.3)$$

where $b = ks$ is the horizontal distance we shall call the base length corresponding to VA_b , the resulting measure of Root Mean Square Vertical Acceleration (RMSVA):

$$VA_b = C \left[\sum_{i=k+1}^{N-k} (S_b)_i^2 / (N - 2k) \right]^{1/2} \quad (2.4)$$

where

N is the number of profile data points and C is a constant required for unit conversion. In this study, $C = 5378 \text{ ft}^2 / \text{sec}^2$ is used to transform units ft / sec^2 assuming profile dimensions are in feet and the hypothetical object is travelling 50 mph.

Specifying the base length b is essential if RMSVA is to be a meaningful description of a road profile. Both theoretically and in practice, one finds that VA_b will tend to increase as b is decreased. Furthermore, in a typical profile, \hat{VA}_b is most sensitive to half wavelengths approximating b . Wavelengths much larger than twice the base length contribute very little to RMSVA and, as a result, their effect on roughness is not revealed. On the other hand, roundoff errors in the computations, or measurements, will ultimately limit the resolution achieved by reducing the base length. In fact, it is this sensitivity of RMSVA to base length that renders it a useful statistic for describing a profile. RMSVA, therefore, should not be regarded as a single roughness index but rather as a set of indices, say VA_i , $i = b_1, b_2, \dots$, which collectively can reveal many of the pavement characteristics usually associated with roughness.

A Comparison of RMSVA With Slope Variance

Before discussing the relationship between MRM and RMSVA, we will first present an argument to the effect that RMSVA can be naturally identified with roughness--more so than the better known statistic, slope variance (SV). First, consider the response of RMSVA to a smooth sine wave of amplitude A and period $T = 2\pi/W$:

$$\begin{aligned}
 VA_b^2 &= \frac{1}{T} \int_0^T \left[A/b^2 (\sin W(t+b) - 2 \sin Wt + \sin W(t-b)) \right]^2 dt \\
 &= \frac{2A^2}{b^4} (1 - \cos Wb)^2 ; \\
 VA_b &= \frac{A\sqrt{2}}{b^2} (1 - \cos Wb) \qquad (2.5)
 \end{aligned}$$

$$\begin{aligned}
&= \frac{2A\sqrt{2}}{b^2} \sin^2 \frac{\pi b}{T} \\
&= 2\sqrt{2} \pi^2 A/T^2 \left[1 - \frac{1}{3} \left(\frac{\pi b}{T} \right)^2 + \frac{2}{45} \left(\frac{\pi b}{T} \right)^4 \dots \right] .
\end{aligned}$$

Equation 2.5 reveals that when base length b is small in comparison to period (or wavelength), VA_b is proportional to A/T^2 . (This limit can also be obtained by differentiating the sine function twice, squaring, and integrating.)

For slope variance (SV) calculated at base length b , we instead have

$$\begin{aligned}
SV_b &= \frac{1}{T} \int_0^T \left[\frac{A}{b} (\sin W(t+b) - \sin Wt) \right]^2 dt \\
&= (A/b)^2 (1 - \cos Wb) \\
&= 2\pi^2 (A/T)^2 \left[1 - \frac{1}{3} \left(\frac{\pi b}{T} \right)^2 + \dots \right] .
\end{aligned} \tag{2.6}$$

Therefore, if SV were the accepted roughness measure, the "roughness" of a sine wave would depend only on its shape (A/T). This insensitivity of SV (for small b) to the scale of disturbance being measured is somewhat contrary to the usual notion of roughness.

An examination of road profiles which had been digitally filtered revealed, as one might expect, that the amplitudes of predominant variations in profiles are approximately proportional to their wavelengths (Fig 2.2). Figure 2.3 illustrates the response of VA_b to wavelength, assuming proportionality of amplitude ($A = T/2\sqrt{2}$), for several baselengths. The corresponding curves for SV would be horizontally displaced, depending on b , but would each approach a maximum of $SV = \pi^2/4$ as T increases.

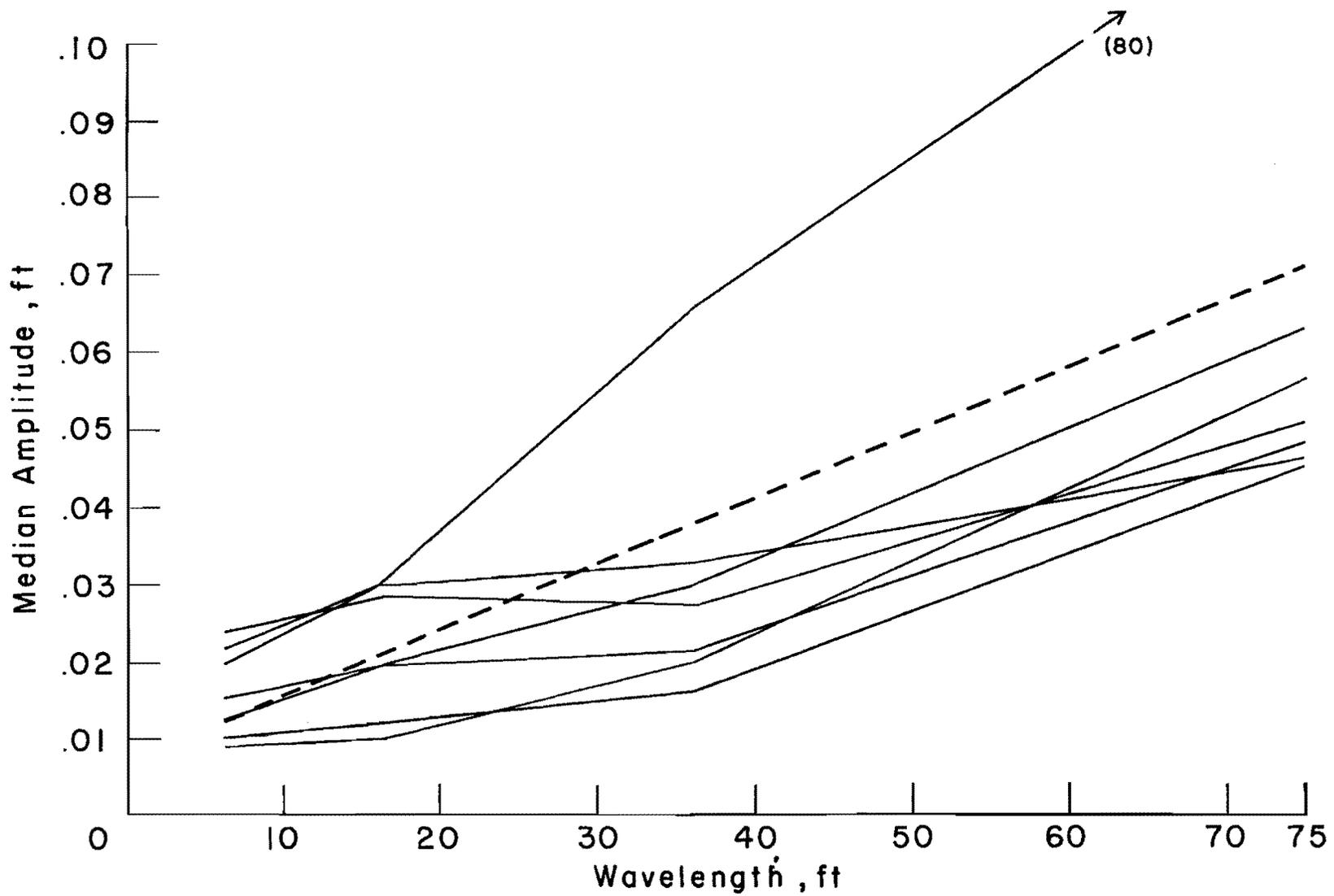


Fig 2.2 Relationship between median wave amplitude and wavelength as revealed by filtered profiles taken from seven ACP road sections near Austin, Texas. The dashed line illustrates a typical wavelength/amplitude ratio at 1000.

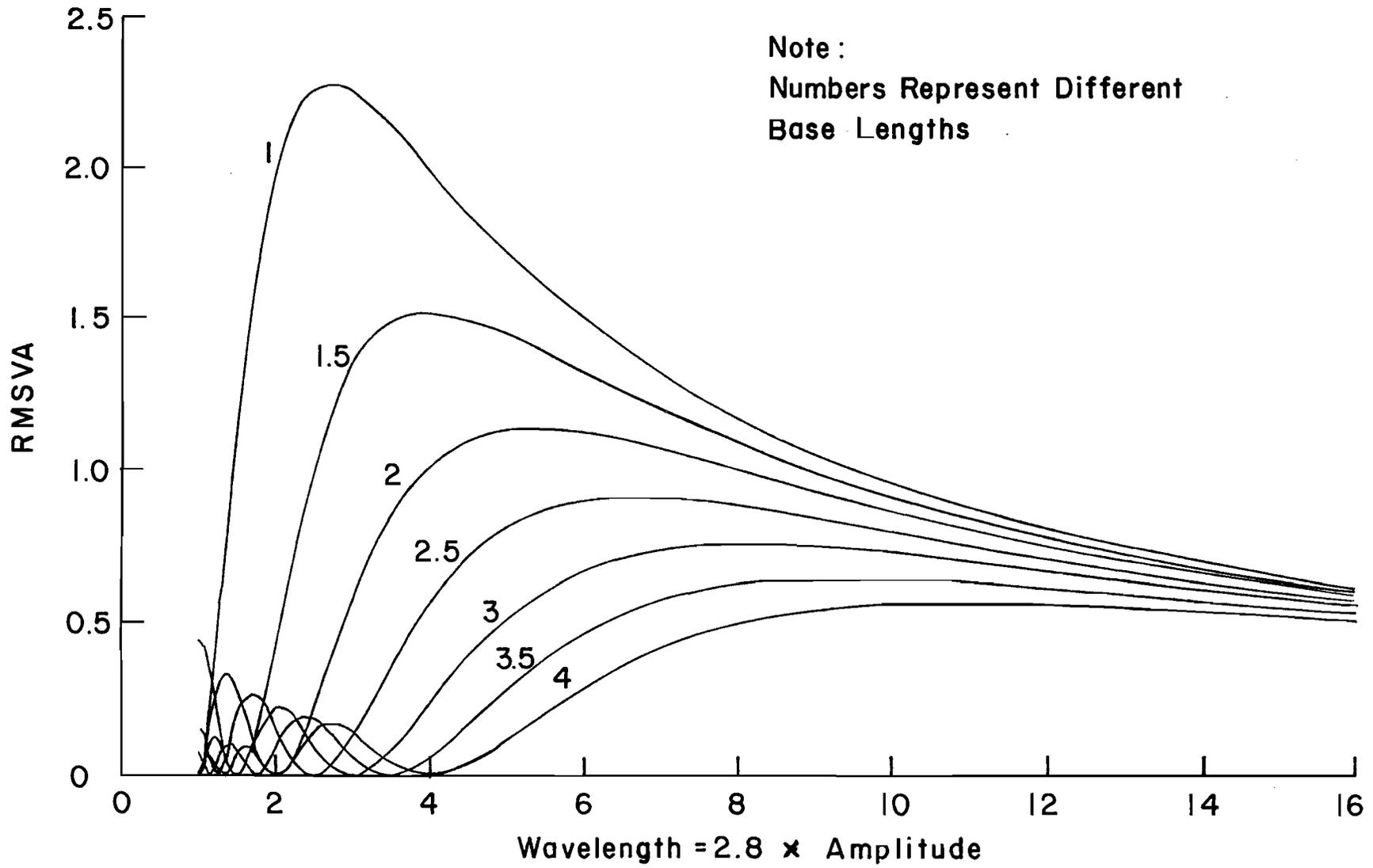


Fig 2.3. Response of RMSVA to a regular waveform at difference base lengths.

Another problem with slope variance is that the mean slope, about which variance is to be measured, is not necessarily zero in practice. Over what length of road should the mean be taken? One would not want the roughness contribution of a short road segment to depend on any (global) property of the entire road being measured. A Maysmeter is a summation device which totals axle-body displacement, a phenomenon that depends only on local roughness characteristics. RMSVA, which is a function of the differences between adjacent slopes, is likewise an averaging, or rather a root-mean-squaring, of local properties of the profile.

Although RMSVA, in a physical sense, could be related to the vertical acceleration of a tire travelling over the road (assuming base length is carefully chosen), this is not really a good characterization of its utility as a roughness measure. The Maysmeter simulation described in Chapter 3 involves only RMSVA at base lengths 4 and 16 feet. It would be better, perhaps, to think of VA_b as the root-mean-square vertical velocity of a front tire with respect to a rear tire when they are travelling b distance apart. In any case, slope variance seems to have no such mechanistic interpretation.

A ROUGHNESS CHARACTERIZATION OF THE AUSTIN TEST SECTIONS

To determine whether or not RMSVA profile indices could be specifically tailored to reveal Maysmeter-type roughness, the computer program VERTAC (Appendix B) was written to compute RMSVA for a selected sequence of base lengths. For our original correlation studies, the following indices were

obtained from an October 1977 measurement of 29 ACP test sections near Austin:

$$VA_{.5}, VA_1, VA_2, VA_4, VA_8, VA_{16}, VA_{32}, VA_{65}$$

The subscripts represent base length in feet, and the units are feet/sec², as would correspond to the RMSVA of a point in contact with the road travelling 50 mph. In each case, a sampling interval of $S = 0.169$ feet was used over a section length of 1050 feet (since $S > 1/6$, the next-to-last base length is 65 feet, not 64 feet).

The small sampling interval was a characteristic of the profilometer data available to us and was not considered essential. In fact, it was found that increasing the sampling interval to 2 feet had little effect on the computed indices for baselengths 2 feet or longer. A small sampling interval simply provides more profile locations to be accounted for in the mean-square.

The above sequence of base lengths was selected in view of both the sampling interval and the observed correlation between indices. For example, it was found that the correlation between VA_2 and VA_4 was approximately the same as that between VA_4 and VA_{16} . Furthermore, the comparison with Maysmeters described in Chapter 3 shows that the sequence encompasses the smaller range of base lengths useful for predicting Maysmeter-type roughness (Fig 3.1).

The index VA_b was computed as the average RMSVA, at base length b , for the right and left wheelpath profiles. As such, the repeatability of the VA_b measurements for successive runs of the profilometer was found to be

excellent. The standard errors of duplication (SE), for example, can be compared with the standard deviation (SD) of the section means (the average of 2 runs) taken over the 29 Austin test sections (Table 2.1).

TABLE 2.1. SUMMARY OF RMSVA MEASUREMENTS ON 29 ACP TEST SECTIONS

	<u>.507</u>	<u>1.014</u>	<u>2.028</u>	<u>4.056</u>	<u>8.112</u>	<u>16.224</u>	<u>32.448</u>	<u>64.849</u>
MEAN VA_b	82.52	28.48	9.13	3.43	1.38	0.64	0.23	0.13
SD	29.71	10.56	3.73	1.77	0.86	0.43	0.19	0.075
SE	2.60	0.83	0.28	0.07	0.13	0.008	0.017	0.005

The RMSVA data for the sections reveal the important fact that variations between sections are an order of magnitude greater than variations between repeat runs. This means that, regardless of their correlations with Maysmeters, PSR, etc., the RMSVA indices represent persistent and distinguishable traits of the road sections. This is not to say that interactions between the road and profilometer system do not play a role. Bounce in the road-following wheels, for example, is believed to affect the short base-length measurements -- particularly for $b = .5$ feet. Such effects have been kept reasonably constant by proper maintenance of the hardware and by operating at a fixed speed of 20 mph.

The Serviceability Index SIV

Although the RMSVA indices are well defined and precisely measurable with the profilometer, they are difficult to interpret by themselves because they tend to increase rapidly in magnitude as base length decreases. The measure of riding quality with which engineers are most familiar is the Serviceability Index (SI), a number between 0 and 5. As will be described in Chapter 3, such an index, SIV, was developed based on RMSVA; however, because it was intended as a provisional calibration standard, it is a function solely of a Maysmeter simulation (inches per mile) comprised of just two RMSVA indices:

$$MO = -20 + 23VA_4 + 58VA_{16} \quad (2.7)$$

where VA_4 and VA_{16} are the mean wheelpath RMSVA at base length 4 feet and 16 feet, respectively. The serviceability estimate is calculated as

$$SIV = 5e^{-\left[\frac{\ln(32MO)}{8.4933}\right] 9.3566} \quad (2.8)$$

and as such is a measure of roughness primarily in the 8-35 feet wavelength range.

SI Values at Different Base Lengths

SIV has proved to be an effective standard for Maysmeter calibration; however, other RMSVA indices (base lengths 2, 4, 8, 16, 32, 65, and 130 feet) are genuine roughness traits which have been useful for comparing pavements in other studies. Therefore, to make such comparisons easier, rescaled versions of these indices are usually provided which resemble a sequence of "SI_b values" in the range 0 to 5. This is accomplished simply by replacing term MO in equation (2.8) with MO_b, a linear function of VA_b obtained by a least-squares fitting of VA_b against MO:

$$\begin{aligned}
 MO_1 &= 16.16 + 2.94 VA_1 \\
 MO_2 &= -28.59 + 13.38 VA_2 \\
 MO_4 &= -23.51 + 34.46 VA_4 \\
 MO_8 &= 6.13 + 66.13 VA_8 \\
 MO_{16} &= 10.83 + 139.18 VA_{16} \\
 MO_{32} &= 10.10 + 296.66 VA_{32} \\
 MO_{65} &= 19.28 + 602.00 VA_{65} \\
 MO_{130} &= 26.30 + 1643.80 VA_{130}
 \end{aligned}$$

The main advantage of such a scaling is that the means and standard deviations of these different SI_b values, as determined on 31 ACP test sections near Austin (April 1981), are approximately the same, making it easier to judge the significance of these measurements on particular pavements. The test sections encompass a variety of roughness conditions, exhibiting a SIV range of .63 to 4.83, with mean 3.12 and standard deviation of 1.23. For assistance in interpreting, or comparing "roughness signatures"

in future studies, the results obtained during April 1980 and April 1981 on the 29 test sections are given in Appendix D.

The use of the term " SI_b " for a scaled RMSVA index is potentially misleading because the index is related to Present Serviceability Rating (PSR) in only a very indirect sense: via its correlation with SIV, which in turn is based on a study of the correlation between the Maysmeter roughness and a PSR prediction equation established during a 1968 human rating experiment. A future rating session, utilizing modern automobiles and an improved profile measuring system, should help determine the relative significance of the different roughness indices in terms of actual riding quality. In the meantime, the RMSVA summary data will be used for comparing pavements and for detecting changes in different components of roughness.

RMSVA data for two sections known to be subject to deterioration from expansive clays are shown in Fig 2.4 (SI_b plotted against base length, b), along with the corresponding values obtained periodically during the previous 18 months (dashed lines). Notice that the spectra of SI_b values form distinctive "signatures" which, in this case, changed very little during the last four-month time period. The test section (lower figure) shows the effect of treatment by a fabric moisture seal sometime prior to the first profilometer run, in June 1979. The differences, however, are confined to the longer RMSVA base lengths and would probably not be noticed in readings from a Maysmeter. Data for these sections were provided by Texas SDHPT Engineer Malcolm Steinberg.

A more typical situation is that of Austin test section No. 23 (see Appendix D), where RMSVA signatures taken one year apart show the effect of an intervening overlay on short-wavelength roughness. In this case,

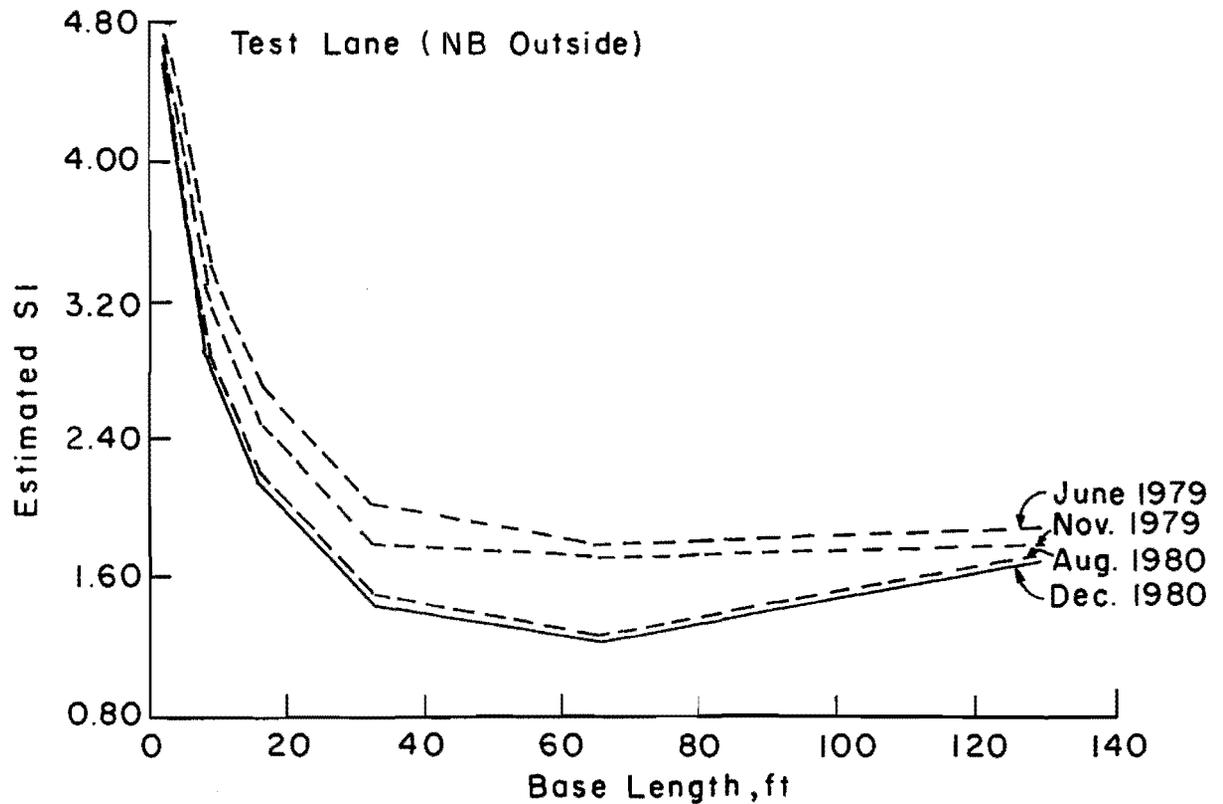
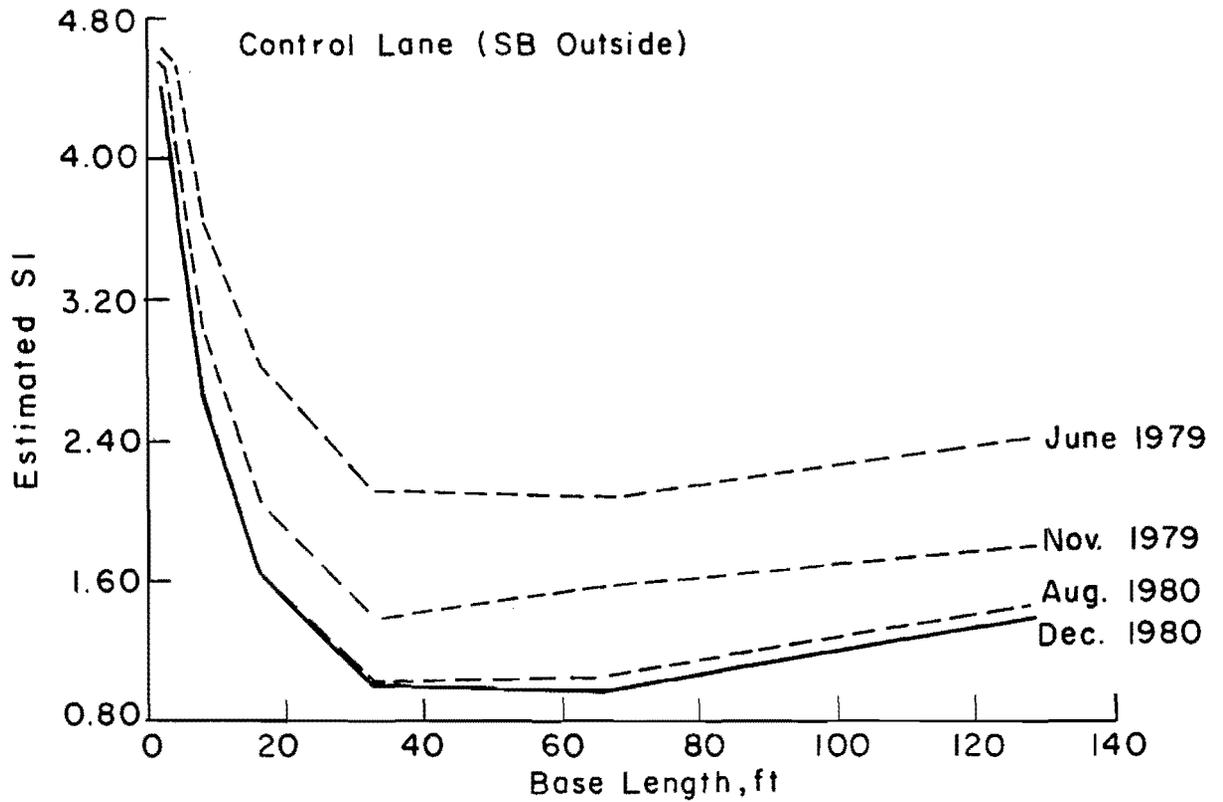


Fig 2.4. RMSVA signatures for untreated (top) and treated (bottom) ACP sections in a swelling clay environment--Loop 410, San Antonio, Texas.

Maysmeter would have detected a distinct increase in serviceability (SIV),
from 2.6 to 4.0.

CHAPTER 3. DEVELOPMENT OF A MAYSMETER CALIBRATION STANDARD

While the SD Profilometer is used in Texas both for calibration and for projects requiring pavement profiles, most roughness monitoring is accomplished by a fleet of Maysmeters. Each vehicle has a chart recording system with a stepper motor that advances the chart in 1/64-inch increments, an increment corresponding to 1/10-inch vertical displacement between the vehicle's body and rear axle. Most of the Maysmeters are trailer-mounted units which are also equipped with digital readouts--32 counts per inch of chart, or five counts per inch of axle displacement. The roughness observations, in either case, are converted to units of inches of axle displacement per mile of roadway travelled. (The units "counts/.2 miles" are sometimes used and can be considered equivalent.) The road sections are normally run at 50 mph.

AN EXAMINATION OF MAYSMETER DATA

For our work in attempting to simulate the Maysmeter with road profile statistics, we had available the calibration session results for 8 devices (Table 3.1). All measurements were obtained within a month's period of an October 1977 profilometer run on 29 test sections near Austin. For convenience, the Maysmeters are labeled M1, M2, . . . , M8, with M1 through

TABLE 3.1. PROFILOMETER AND MAYSMETER DATA USED IN THE DERIVATION OF RMSVA MAYSMETER SIMULATION MØ.

SEC	RMSVA INDICES						MAYS METER MEASUREMENTS								
	VA2	VA4	VAS	VA16	VA32	MO	M1	M2	M3	M4	M5	M6	M7	M8	
1	7.75	2.56	1.01	.52	.30	69.0	*	---	59.0	67.5	63.0	60.0	112.5	113.8	116.0
2	11.93	5.16	2.27	.95	.46	153.8	*	173.5	144.7	185.5	145.2	170.8	244.7	249.0	309.6
5	6.28	2.53	1.00	.48	.22	66.0	*	72.3	60.3	56.0	56.8	59.3	112.3	96.8	113.6
7	3.00	1.06	.52	.30	.16	21.8	*	33.3	38.0	31.0	22.3	20.0	52.3	44.8	44.8
8	6.16	2.24	.99	.41	.17	55.3	*	65.0	64.0	60.0	53.8	59.3	92.5	92.5	88.0
9	3.90	1.45	.64	.30	.16	30.7	*	44.5	45.8	52.3	50.8	37.8	73.3	---	134.4
10	4.00	1.15	.40	.22	.14	19.2	*	29.0	38.0	24.3	18.0	22.0	53.3	43.3	36.8
11	5.83	1.75	.52	.20	.11	31.9	*	24.0	40.8	24.3	32.5	24.0	66.5	55.0	40.0
12	9.62	3.13	.95	.48	.30	79.8	*	72.0	84.5	75.5	70.0	85.3	156.3	146.5	134.4
13	11.91	3.78	1.18	.50	.23	95.9	*	98.5	100.8	113.0	98.3	122.0	211.0	189.7	187.2
14	6.00	2.01	.80	.39	.22	48.9	*	44.8	43.5	49.3	55.0	18.0	94.0	89.0	72.8
15	11.13	3.11	.93	.32	.16	70.1	*	64.8	75.5	83.0	57.5	72.3	149.8	94.0	103.2
19	6.41	2.16	.91	.53	.26	60.4	*	56.8	58.0	55.8	50.3	56.5	116.5	99.5	110.4
23	9.37	3.38	1.18	.48	.17	85.6	*	---	75.5	78.3	77.3	80.3	149.0	139.5	164.8
28	4.04	1.50	.75	.42	.19	38.9	*	---	43.5	41.3	44.8	38.8	80.0	72.8	73.6
32	9.40	2.82	.94	.37	.22	66.3	*	51.3	52.8	50.0	43.8	37.0	117.3	84.5	84.0
33	7.61	3.37	1.50	.65	.26	95.2	*	102.2	88.3	118.2	98.3	106.0	162.0	125.5	165.6
34	8.91	3.33	1.25	.50	.23	85.6	*	89.5	81.8	94.0	74.3	85.0	142.5	143.8	144.0
35	14.37	6.58	2.20	.98	.39	188.2	*	198.7	178.0	195.8	164.3	201.8	---	---	406.4
36	6.58	1.78	.55	.27	.08	36.6	*	25.8	24.3	22.3	19.3	24.5	95.0	43.3	35.2
37	9.84	3.18	1.03	.41	.21	76.9	*	87.5	84.8	74.8	70.8	76.5	168.5	146.5	---
38	14.53	4.28	1.17	.45	.22	104.5	*	112.7	114.5	142.0	117.8	113.5	---	---	221.6
39	11.77	5.08	2.35	.92	.37	150.2	*	143.5	138.5	159.0	131.8	130.7	223.0	220.5	279.2
40	7.67	2.34	.83	.45	.29	59.9	*	64.5	63.5	71.8	67.3	50.3	93.0	100.2	93.6
41	11.92	4.37	1.74	.77	.30	125.2	*	139.7	123.2	133.0	121.3	133.3	194.5	182.7	207.2
42	16.63	7.74	3.14	1.23	.44	229.4	*	229.8	212.5	235.5	224.5	230.7	---	---	368.8
43	16.91	6.55	3.07	1.45	.71	214.7	*	223.7	216.8	227.2	218.0	223.5	---	---	404.8
44	10.18	4.88	2.73	1.67	.85	189.1	*	202.0	179.0	192.3	154.0	201.0	---	---	376.6
45	11.19	6.09	3.44	1.86	.83	227.9	*	259.2	188.8	241.8	223.0	241.5	---	---	419.2

M5 denoting the trailer-mounted units. Each table entry was obtained by averaging the results of four runs on a 0.2-mile section. (A fifth run, which had deviated most from the overall mean, was first excluded.) This redundancy provides a measure of the repeatability of the Maysmeter for successive runs. Table 3.2 contains the average section mean, the standard deviation of the section means, SD, and the standard error of repeatability, SE, for each unit. The repeatability (although underestimated slightly because of the excluded value) is quite good considering that 100 ± 5 in./mile corresponds roughly to a Serviceability Index of 2.7 ± 0.1 .

TABLE 3.2. MRM RESULTS - OCT - DEC 1977

	M1	M2	M3	M4	M5	M6	M7	M8
MEAN	104.17	93.73	101.87	90.46	95.91	128.67	116.96	174.02
SD	70.13	56.02	67.83	60.05	69.16	54.17	56.34	125.27
SE	5.01	3.88	6.47	3.94	7.19	6.33	7.27	16.05

Most relevant to the problem of calibration are the relationships shown to exist among the different units. The correlation matrix for the data in Table 3.1 indicates that the Maysmeter roughness readings are highly correlated and, in fact, plots show that the relationships are linear. If we were to seek a simple linear calibration function, with one of the units selected as the reference device, then a good choice of reference would be

unit M3, whose measurements explain a high percentage of the section-to-section response variation in the other units. Specifically, the coefficients of determination (R^2) and the standard errors of estimate (SE) for the model

$$M_i = \alpha_i + \beta_i M_3 + \epsilon_i \quad (3.1)$$

are:

	<u>M1</u>	<u>M2</u>	<u>M4</u>	<u>M5</u>	<u>M6</u>	<u>M7</u>	<u>M8</u>
R	.975	.970	.976	.974	.857	.900	.964
SE	11.3	10.0	9.4	11.4	21.0	18.3	24.3

Each of the other trailer units would serve about equally well as a standard reference. The car units M6, M7, and M8 behave less predictably (there are no measurements on the rough sections for the first two), although trailer unit M3 explains their response about as well as any one car would explain the other two cars.

Such close agreement between the different units over a two-month period convinced us that a linear calibration model such as Eq 3.1 would be adequate provided we could find a profile statistic that could effectively assume the role of M3. Unlike M3, of course, it should also have long-term stability, depending only on the profilometer or other instruments to obtain a reasonably accurate profile. Moreover, since calibration requires that measurements from all units be transferable to a common scale, we cannot hope

to find a single statistic that agrees much better with the units than they do between themselves; hence, we can be satisfied if our candidate statistic performs about as well as M3 above. This means that standard errors of estimate (SE) should be on the order of 10 in./mile or 0.2 units of serviceability. This would be a significant improvement, incidentally, over the 0.5 or 0.6 unit SE typically obtained in a calibration based on the serviceability index (SI), a method once used by the Texas State Department of Highways and Public Transportation (Eq 2.1).

CORRELATION OF THE MAYSMETER WITH RMSVA

A comparison of Maysmeter data with the RMSVA indices of Table 2.1 reveals significant individual correlations with the subset VA_2 , VA_4 , VA_8 , VA_{16} , and VA_{32} (Table 3.3), and plots show that the relationships are essentially linear. Also, the replication errors for the Maysmeters do not seem to be significantly larger for the rougher sections. Therefore, no transformation or weighting was considered necessary when forming the regression models.

If we were to choose a single RMSVA index to explain the Maysmeter results then it would be VA_8 . For example, when the San Angelo Unit M1 is regressed on VA_8 we obtain

$$\begin{aligned} M1 &= 5.1 + 76.7 VA_8, \\ R^2 &= .964, \\ SE &= 13.6. \end{aligned}$$

TABLE 3.3. INDIVIDUAL CORRELATIONS BETWEEN MRM AND RMSVA INDICES

<u>RMSVA</u>	<u>M1</u>	<u>M2</u>	<u>M3</u>	<u>M4</u>	<u>M5</u>	<u>M6</u>	<u>M7</u>	<u>M8</u>
VA2	0.794	0.858	0.841	0.834	0.829	0.917	0.843	0.820
VA4	0.953	0.970	0.964	0.962	0.960	0.971	0.954	0.955
VA8	0.982	0.963	0.968	0.971	0.964	0.889	0.917	0.961
VA16	0.945	0.904	0.909	0.911	0.920	0.831	0.887	0.924
VA32	0.879	0.840	0.839	0.836	0.848	0.721	0.820	0.862
MO	0.991	0.986	0.984	0.983	0.986	0.955	0.962	0.984

Given the similar results for M2 - M8 (M6 correlates better with VA_4 than with VA_8) we find that VA alone is a much better predictor of Maysmeter response than the serviceability index used previously for calibration.

On the other hand, plots of $VA_{.5}$ and VA_{65} versus Mi showed little or no correlation between Maysmeters and RMSVA at those base lengths. Although $VA_{.5}$ and VA_{65} are genuine characteristics of the sections, as Table 2.1 shows, the Maysmeters are simply not sensitive to those characteristics. (Actually, $VA_{.5}$ is more an interaction between the road and measuring system, in which bounce of the profilometer's following wheels plays a role.)

When multiple regression procedures were applied to the Maysmeter data, it was found that two indices, VA_4 and VA_{16} , were sufficient to explain quite well the variations between Maysmeters on the 29 test sections. Furthermore, no significant improvement in the correlations came about by allowing different combinations, or functions, of RMSVA indices. Figure 3.1 shows that the correlations of the two indices VA_4 and VA_{16} with Maysmeter roughness are large compared with their correlation with each other; hence, each statistic contains relevant information that is not contained in both statistics. Such plots actually indicate that the peak response for most Maysmeter units is at a base length smaller than 8 feet and that another pair, say VA_3 and VA_{12} , might have provided marginally better correlations.

Comparison of the Maysmeter data with VA_4 and VA_{16} produced the following regression equations:

$$\begin{aligned}
 M1 &= -20.1 + 20.9VA_4 + 77VA_{16}, & R^2 &= .98 \\
 M2 &= -9.4 + 22.4VA_4 + 42.2VA_{16}, & R^2 &= .97 \\
 M3 &= -22.1 + 25.9VA_4 + 55.3VA_{16}, & R^2 &= .97 \\
 M4 &= -18.9 + 22.4VA_4 + 63.9VA_{16}, & R^2 &= .97 \\
 M5 &= -29.4 + 24.7VA_4 + 63.9VA_{16}, & R^2 &= .97
 \end{aligned}$$

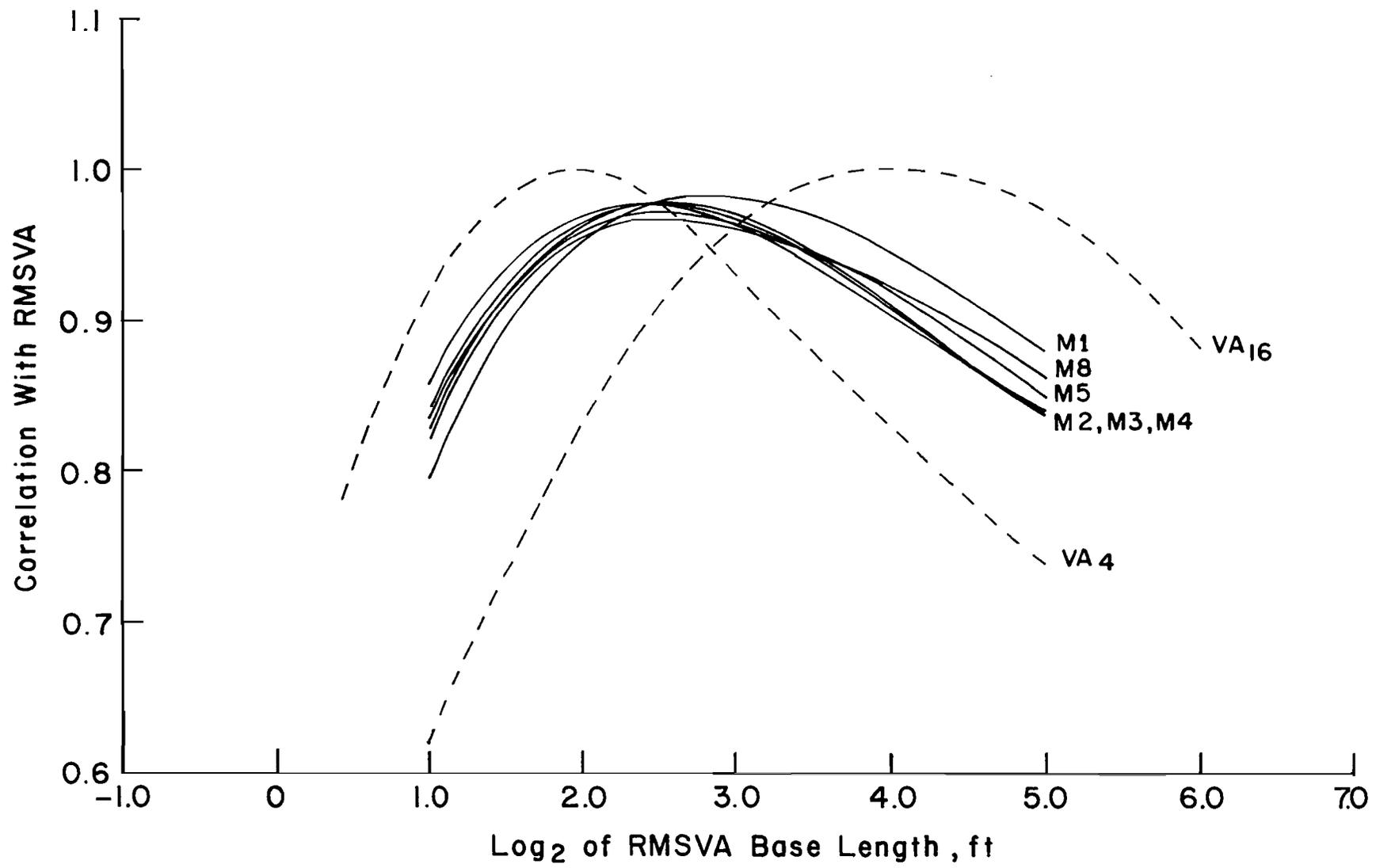


Fig 3.1. Correlation of Texas Maysmeter measurements with RMSVA.

$$\begin{aligned}
 M6 &= 2.5 + 48.1VA_4 - 12.8VA_{16}, & R^2 &= .94 \\
 M7 &= -20.8 + 36.8VA_4 + 71.8VA_{16}, & R^2 &= .93 \\
 M8 &= -41.7 + 40.2VA_4 + 123.9VA_{16}, & R^2 &= .97
 \end{aligned}$$

The quality of fit is illustrated by the plotted results for M1 in Fig 3.2. The equations for the trailers are strikingly similar, whereas the three car-mounted Maysmeters comprise a more divergent group. The decision was made to tailor the calibration standard to the more numerous trailers, because of their superior performance and the likelihood that such a standard would serve the cars about as well as one developed specifically for them.

One might suppose that a calibration procedure could be based on an equation of the form

$$M_i = \alpha_i + \beta_i VA + \gamma_i VA_{16} \quad (3.2)$$

with α_i , β_i , and γ_i serving as the parameters to be adjusted for unit i . This might allow adjustments for more than one operating speed. (A reduced speed, for example, would lessen the relative importance of the longer base length.) Our goal, however, is not to simulate a particular Maysmeter run at an arbitrary speed, but to obtain a calibration model that can serve to scale different units so they are compatible with an accepted standard. Such models cannot contain two explicit profile variables, for the simple reason that, in converting a particular unit's measurement to what a "standard device" would obtain, we would face the impossibility of deriving two unknowns, in this case VA_4 and VA_{16} , from a single quantity.

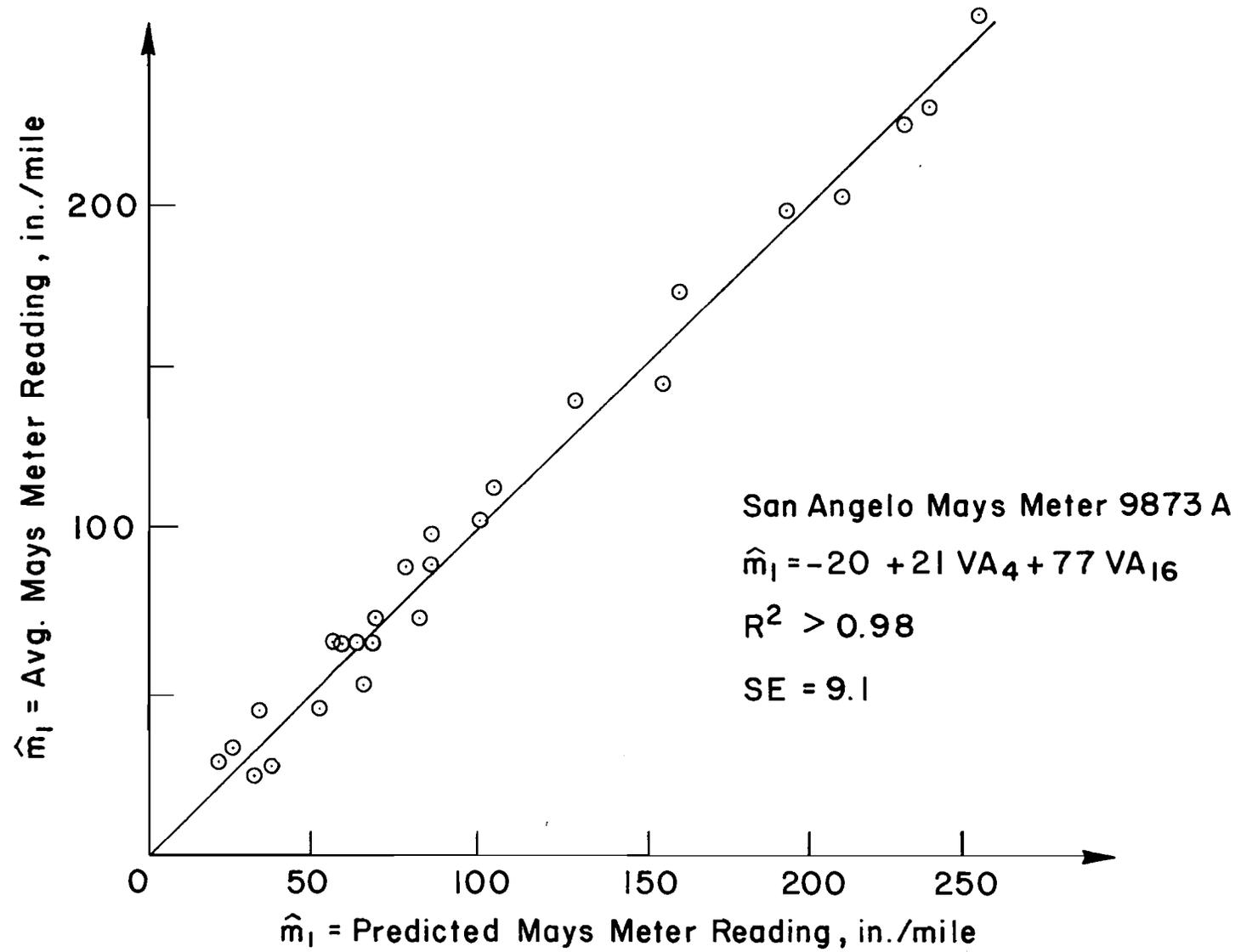


Fig 3.2. A prediction equation for the San Angelo Maysmeter based on RMSVA statistics VA_4 and VA_{16} .

The method we chose, which is justified by the marked similarity of the above regressions, is to fit M_i (run at the standard speed of 50 mph) to

$$M_i = \alpha_i + \beta_i (VA_{4i} + R VA_{16i}) \quad (3.3)$$

where R is predetermined to provide an optimum calibration for the collection of Maysmeter trailers as a whole. This nonlinear regression problem is easily solved by plotting the total regression sum of squares for Eq 3.3 at various values of R and interpolating the minimum. In this manner, $R \approx 2.5$ was obtained.

Such considerations led us, finally, to the calibration model

$$M_i = \alpha_i + \beta_i MO + E_i \quad (3.4)$$

where

$$MO = -20 + 23VA_{4i} + 58VA_{16i} \quad (3.5)$$

The coefficients in the RMSVA statistic, MO , were selected so that α_i and β_i are approximately 0 and 1, respectively, for the Maysmeter trailers. Thus, MO will serve as our simulated Maysmeter. The results of fitting model 3.4 to the Maysmeter and RMSVA data (Table 3.1) are given in Table 3.4.

TABLE 3.4. REGRESSION RESULTING FROM FITTING EIGHT MAYS METERS TO THE LINEAR MODEL OF EQUATION 9

	M1	M2	M3	M4	M5	M6	M7	M8
α_i	-1.7	9.6	0.2	0.5	-7.9	28.6	8.9	-11.9
β_i	1.07	0.88	1.06	0.94	1.08	1.42	1.49	1.93
R^2	.981	.972	.969	.967	.972	.913	.925	.968
SE	9.7	9.5	12.2	11.1	11.9	16.4	15.8	22.8

A comparison of these results with those of the 3-term model (3.2) shows that little is lost by adopting a fixed combination of VA_4 and VA_{16} , even if our purpose were not calibration but actual prediction of a particular Maysmeter's behavior. More important, M0 is at least as successful in describing the Maysmeter data as is any one of the units themselves, including M3, which it most closely resembles. (This conclusion is supported not only by data used in the development of M0 but by subsequent analyses of new data.) Furthermore, when plots of M_i versus M0 are compared, the scatters about the regression lines are dissimilar; that is, what is not explained by M0 is also not explained by another Maysmeter in terms of linear relationships.

The regression results of Table 3.4 are illustrated graphically in Fig 3.3, which shows the relationships among the eight Maysmeters we examined. Although the conclusion is based on less data, the car-mounted Maysmeters obviously differ from the trailers in their relationship to M0. The five trailers, however, are so similar in their response that they would seem to be indistinguishable and thus be in no need of calibration. Yet,

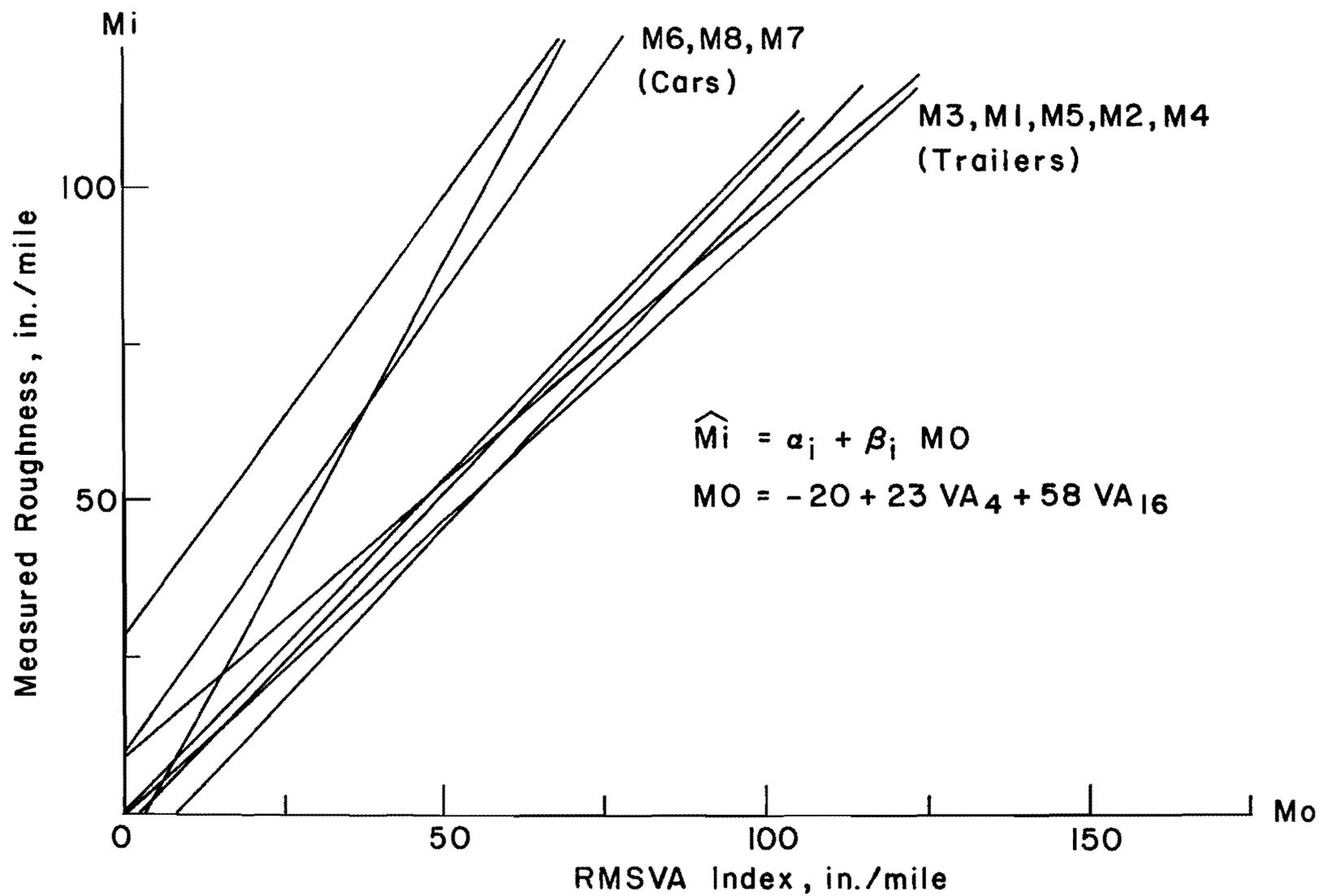


Fig 3.3. Calibration session results for eight Texas Maysmeters.

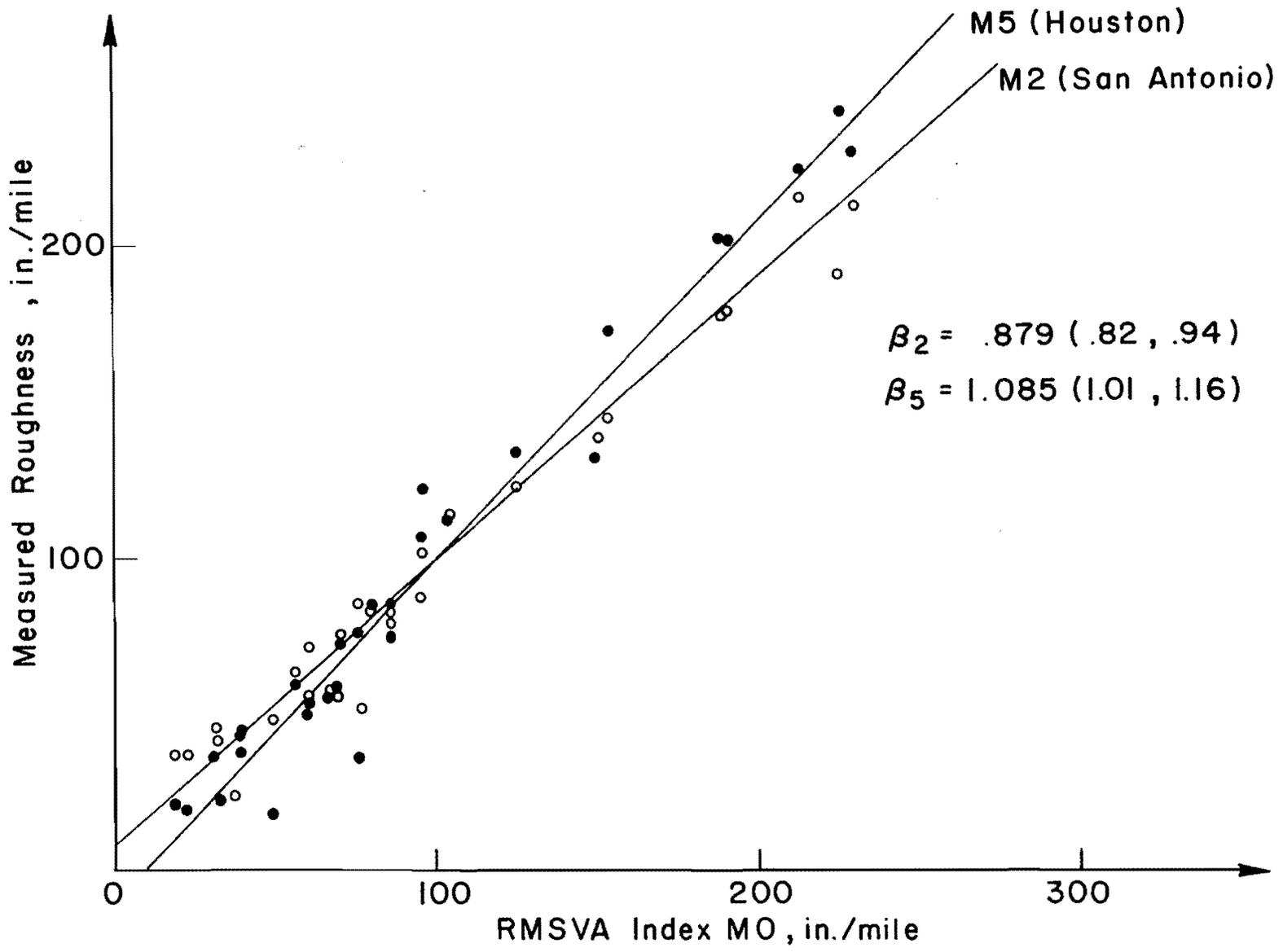


Fig 3.4. Calibration results for two trailer-mounted Mays Meters: 95% confidence intervals for calibration parameters β_2 and β_5 are given.

their correlation with M0 is strong enough that units as similar to each other as M2 and M5 can be separated, as is shown in Fig 3.4. Ninety-five percent confidence intervals for their slope parameters (B_2 and B_5) do not, in fact, overlap.

Several of the pavement test sections we examined had one wheelpath with a profile significantly rougher than that of the other wheelpath. It is important that measurements of both wheelpaths be obtained, for we found that the procedure of averaging the RMSVA indices from two wheelpaths resulted in an index that correlated much better with the Maysmeters than did any single-wheelpath index. We believe that simply averaging the two wheelpath indices is preferable to including a summary statistic based, for example, on the "difference profile," for the latter method would make it difficult to use one wheelpath profile when that of the other wheelpath, for some reason, is not available. The discrepancy between wheelpaths is usually much smaller than that between different road sections, and, when exceptions have occurred, a simple average usually has succeeded in aligning with the calibrated Maysmeter's results.

We summarize the Maysmeter/RMSVA correlation study as follows: a profile statistic based on RMSVA at base lengths 4 ft and 16 ft was successful in explaining approximately 97 percent of the response variation between five trailer-mounted Maysmeters on 29 ACP pavement test sections. This corresponds to a prediction standard error of about 10 percent of the Maysmeter reading (inches/mile), which compares favorably with what would be achieved if an actual Maysmeter were singled out as the reference device (Eq 3.1). Results for the three Maysmeter cars were not quite as favorable ($R^2 = .91, .93, \text{ and } .95$); however, data for two of the cars were incomplete.

The correlation studies which produced the profile statistic M_0 (Eq 3.5) were carried out in early 1978 and since then the statistic has been used regularly by the Texas State Department of Highways and Public Transportation for calibrating their Maysmeters. Although M_0 was "tailored" to describe Maysmeter data obtained in November 1977, subsequent Maysmeter calibrations have continued to demonstrate the high correlations described above.

The Linear Calibration Model

The general problem of calibration in which one has N pairs of observations, say (x_i, y_i) , $i = 1, \dots, N$, where x_i is precise and expensive to obtain while y_i is relatively imprecise and less expensive to obtain, has been given considerable recent attention by statisticians (see, for example, Scheffe, Ref 15). Much of the discussion has dealt with the relative merits of Classical linear calibration, where one fits the N observation pairs to the model

$$y_k = \alpha + \beta x_k + \epsilon_k, \quad k = 1, \dots, N \quad (3.6)$$

and estimates a future x by

$$\hat{x} = (y - \hat{\alpha}) / \hat{\beta},$$

and inverse linear regression, where one assumes

$$x_k = a - by_k + e_k, \quad K = 1, \dots, N \quad (3.7)$$

and regresses x on y to obtain the prediction equation

$$x = \hat{a} + \hat{b}y .$$

We have chosen the former method, primarily because it is customary to regard the independent variables in a regression model as being precisely known and usually causal to the dependent variable or response. Although we will eventually convert a calibrated unit's measurement M_i to a serviceability estimate by way of

$$\hat{M}_i = (M_i - \alpha_i) / \beta_i$$

the fact that β_i is approximately one tends to minimize the distinction between the classical and inverse approaches.

Furthermore, it is advantageous to obtain precise calibration parameters α_i and β_i based on model 3.6 if one is interested in detecting physical changes in units. The calibration regression lines for the eight Maysmeters (Fig 3.3) might suggest that the trailer units, M1 - M5, would be nearly indistinguishable. Yet, the quality of the model is such that units quite similar to each other can be separated (Fig 3.4) and the effects of wear or parts replacement in single units can be detected.

TABLE 3.5. SERVICEABILITY RATING OF AUSTIN TEST SECTIONS

SEC	SI2	SIV
1	3.20 (.08)	2.64 (.01)
2	1.14 (.15)	1.16 (.01)
3	3.77 (.04)	3.75 (.06)
5	4.28 (.13)	4.66 (.02)
6	3.17 (.03)	2.59 (.09)
7	4.56 (.27)	4.83 (.01)
8	3.75 (.01)	3.54 (.02)
9	3.31 (.02)	3.15 (.01)
10	4.71 (.12)	4.46 (.00)
12	3.57 (.38)	2.25 (.07)
13	3.19 (.02)	2.50 (.01)
14	3.53 (.26)	3.42 (.08)
15	3.92 (.13)	3.69 (.00)
19	3.84 (.04)	3.42 (.02)
23	3.21 (.02)	2.64 (.01)
28	4.09 (.09)	4.12 (.02)
35	1.37 (.05)	1.11 (.00)
36	4.43 (.02)	4.33 (.02)
37	3.19 (.25)	2.45 (.08)
38	3.21 (.06)	2.03 (.01)
39	.41 (.16)	1.03 (.00)
40	3.41 (.11)	3.56 (.02)
41	2.01 (.01)	2.07 (.00)
43	2.06 (.01)	1.71 (.01)
44	2.57 (.02)	1.18 (.00)
45	.65 (.15)	.38 (.00)
MEANS:	3.10	2.79
SD:	1.17	1.24
REPEATABILITY:	.198	.049

Field Performance

The RMSVA statistic M0 has yet to be employed directly for calibration of Texas Maysmeters. However, for the year beginning April 1978 two types of profile-derived serviceability estimates, SI2 and SIV, were supplied quarterly for use in the nonlinear calibration model (Eq 2.1). SI2 is the index based on power spectral estimates while SIV is derived by submitting the simulation M0 to the same nonlinear regression on SI2, as if it were actual Maysmeter. Hence, SIV is simply a rescaling of M0 so that its values resemble a serviceability index ranging from 0 to 5. The profilometer results for April 1979, which are typical, are summarized in Table 3.5. The standard errors of repeatability, as determined from the pairs of runs, are .16 and .05 for SI2 and SIV, respectively. The overall means and standard deviations for the section measurements, however, are essentially the same for both indices.

A plot of the Houston unit's measurements (M5) versus SI2 (Fig 3.5), shows that the calibration method introduced in 1973 had deteriorated to the extent that a nonlinear relationship between SI2 and M5 could not be detected. Figure 3.6 shows the effect of simply replacing SI2 with SIV. Such improvement has been typical of the calibrations that have taken place since early 1978. Figures 3.7 and 3.8 show results for the same unit when calibrated on a somewhat changed set of sections about 15 months later. In this case, SI (predicted from the regression) is plotted against each of SI2 and SIV. Figure 3.7 clearly reveals the relative insensitivity of SI2 to properties of the rougher pavements. Again, replacement of SI2 with SIV (Fig 3.8) improves the fit dramatically and results in a more effective calibration.

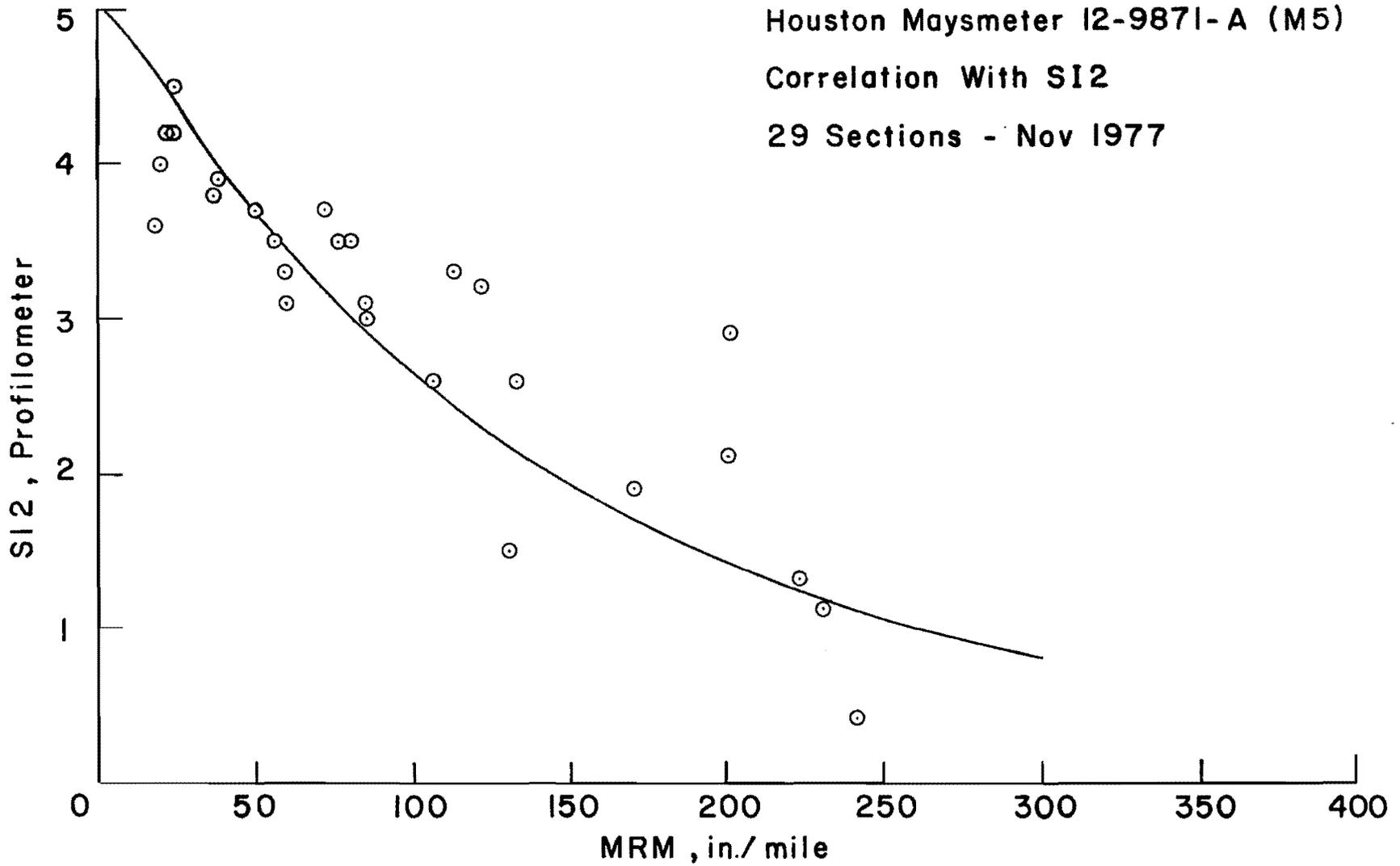


Fig 3.5. Calibration session results for Maysmeter M5 with statistic SI2 employed as the standard.

Houston Maysmeter 12-9871-A (M5)
Correlation With SIV
29 Sections - Nov 1977

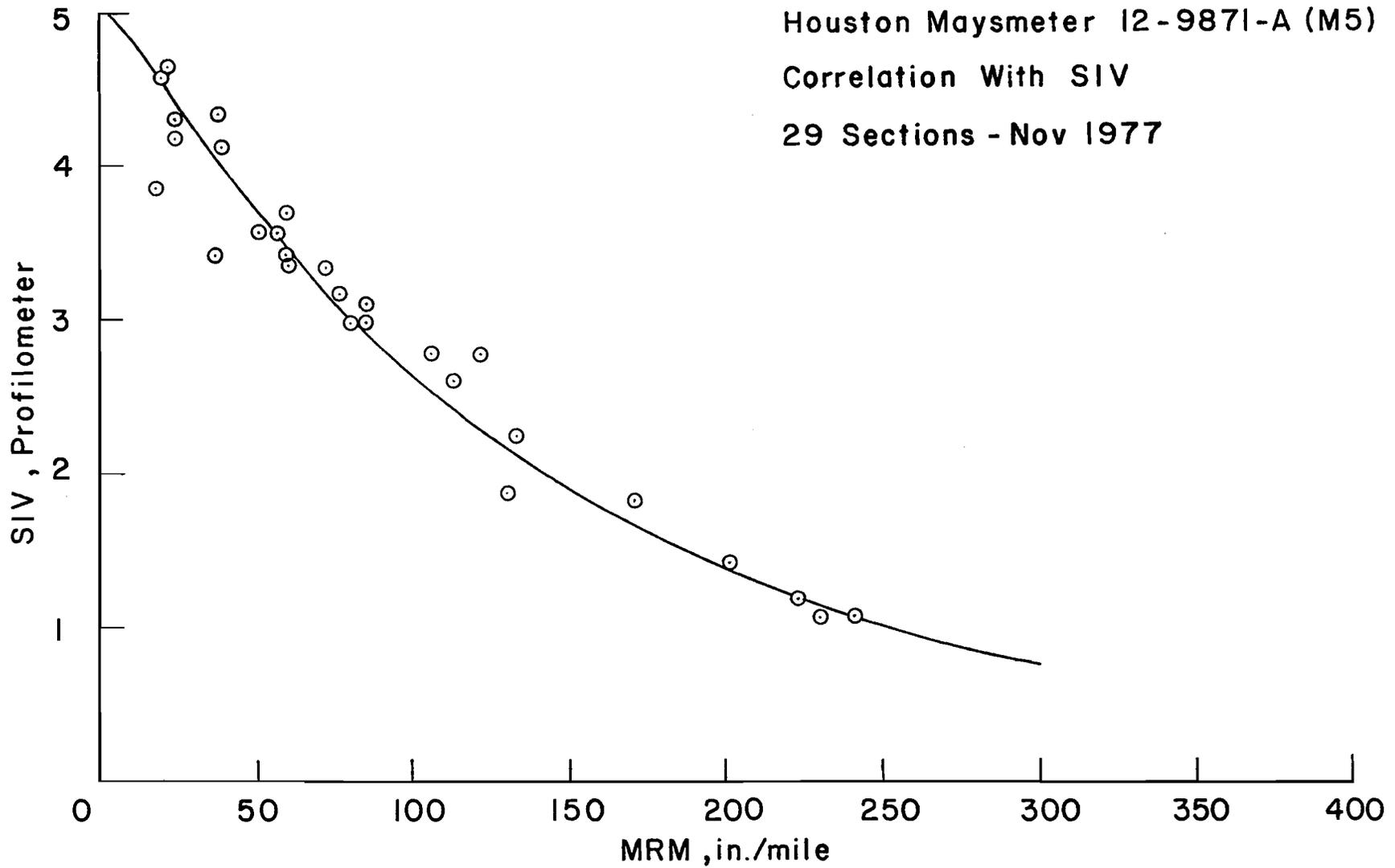


Fig 3.6. Calibration session results for Maysmeter M5 with RMSVA statistic SIV employed as the standard.

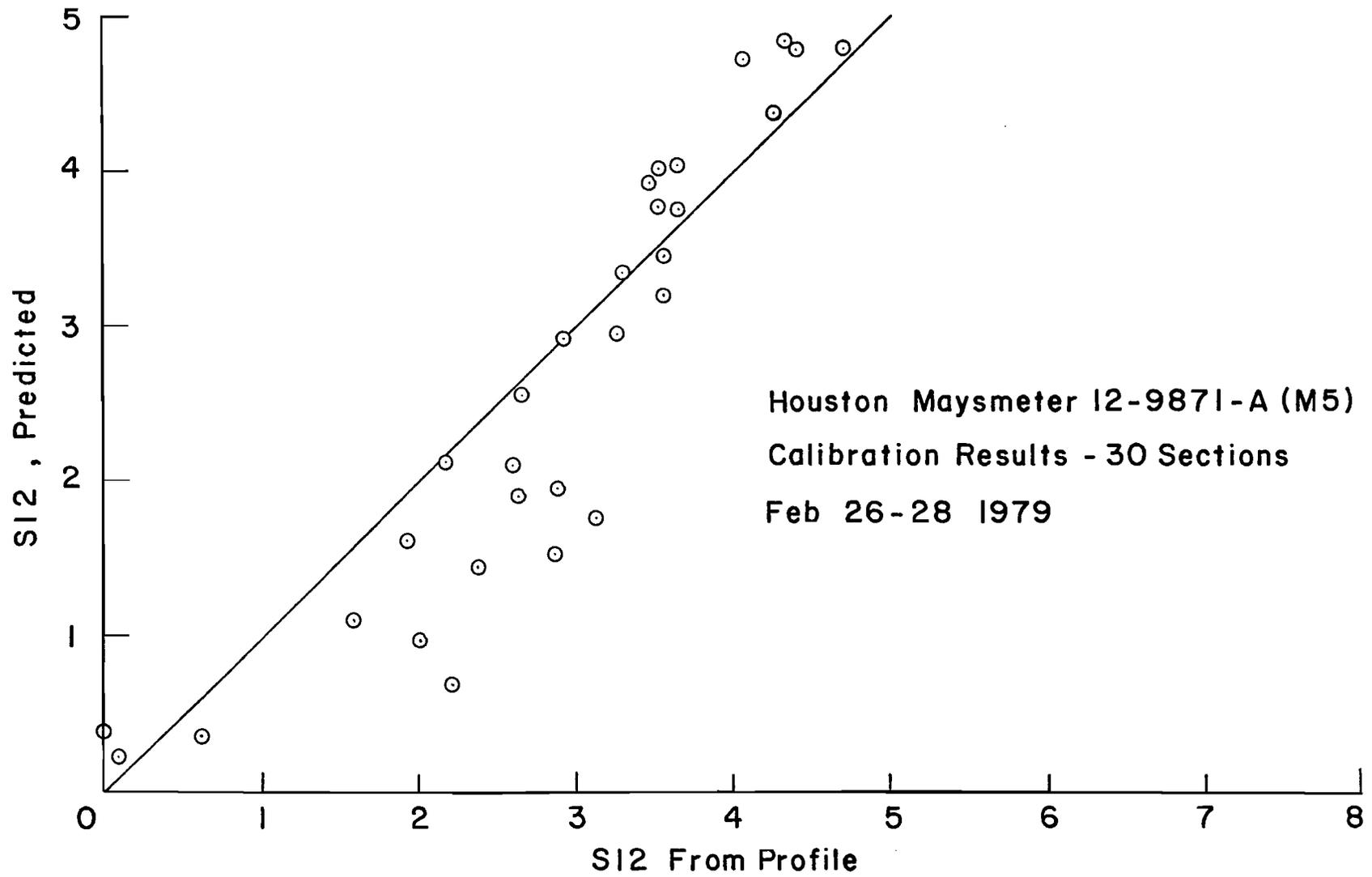


Fig 3.7. Residual plot of calibration session results for Maysmeter M5 with statistic SI2 employed as the standard.

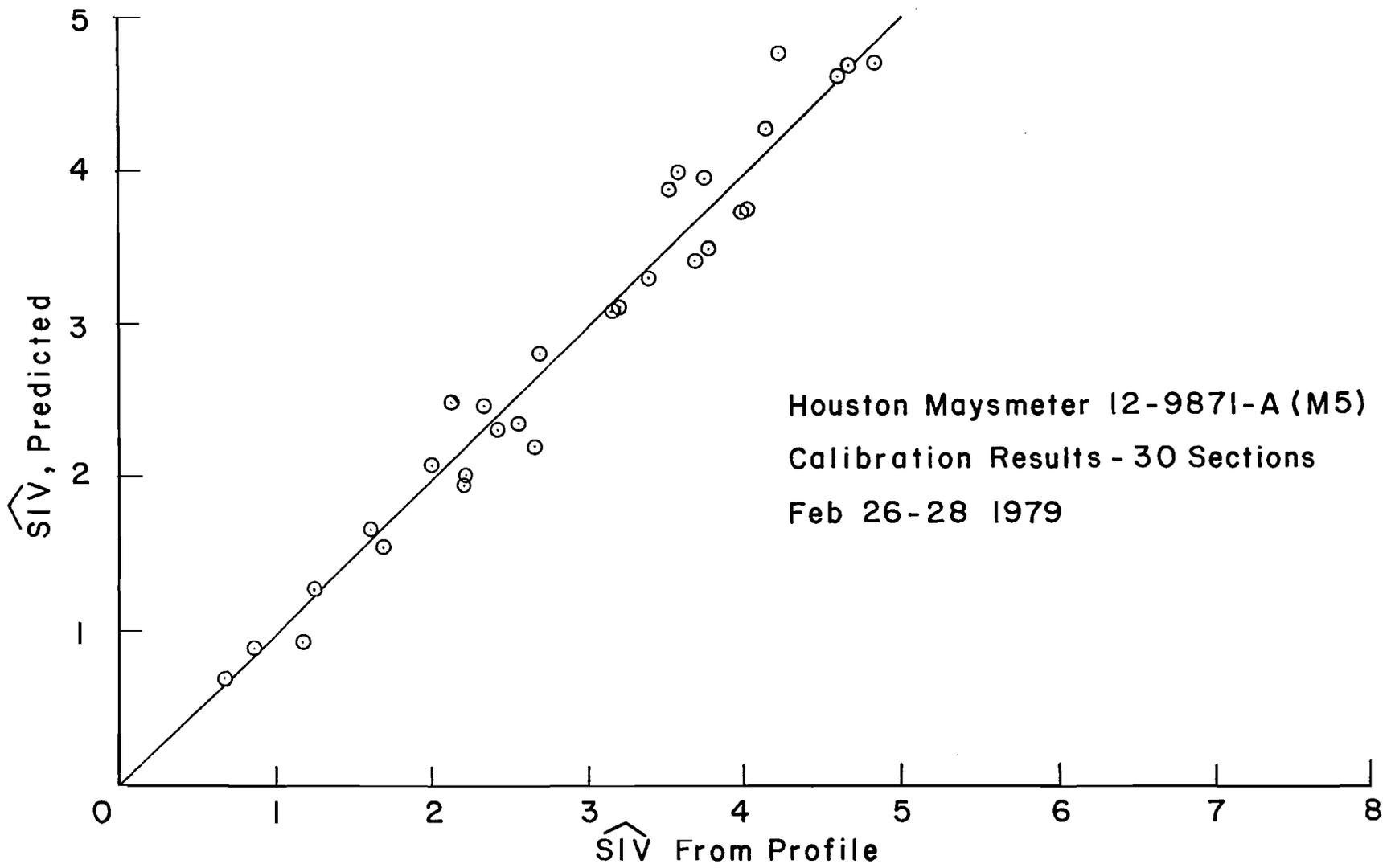


Fig 3.8. Residual plot of calibration session results for Maysmeter M5 with RMSVA statistic SIV employed as the standard.

CHAPTER 4. SUMMARY, CONCLUSIONS, AND RECOMMENDATIONS

RECOMMENDATIONS REGARDING THE TEXAS MRM CALIBRATION PROCEDURE

In Chapter 3, we outlined the development of a road profile statistic, M_0 , which effectively models the Texas Maysmeters. A particular Maysmeter unit i is thus described simply by

$$M_i = \alpha_i + \beta_i M_0 + e \quad (4.1)$$

where M_i is its reading (inches/mile), M_0 is the (RMSVA) profile statistic, and α_i and β_i are parameters (approximately 0 and 1, respectively, for the Maysmeter trailers). The random error e has a standard deviation of approximately 10 in./mile, which is comparable to Maysmeter repeatability for successive runs and is about 15 percent of the standard deviation of M_0 over the Austin test sections.

It is recommended that the SDHPT procedure for Maysmeter calibration be fully revised to incorporate the improved calibration standard. Outlined on the next page are (a) the old method, which employs the original version of calibration regression program MRM CAL, (b) the recommended method, which entails use of the revised program MRM CAL2 and a change in the form of the equation that converts M_i to a serviceability index, and (c) the

provisional method, which entails no changes in procedure but requires simply the replacement of the old standard SI2 with the M0-derived SIV.

Old Method

When a Maysmeter unit is calibrated on the test sections, the nonlinear regression program MRM CAL computes α_i and β_i such that M_i can be converted to serviceability indices $\widehat{SI2}$ by the equation

$$\widehat{SI2} = 5e^{-\left[\ln(32M_i)/\beta_i\right]^{\alpha_i}} \quad (4.2)$$

M_i is scaled by 32 because MRM CAL expects input in units of chart inches. The program was never revised to accept counts, or in./mile, directly because it was felt that the adjustment of parameters α_i and β_i would account for the different kinds of units. The standard error of this estimate (from the regression on profilometer SI2 values) is typically 0.5 or greater. In addition to α_i and β_i , MRM CAL provides a calibration table relating SI2 to M_i .

Recommended Method

The revised calibration program, MRM CAL2, performs the simple linear regression as indicated in Eq 4.1 and calculates parameters and for converting M_i to the "standard MRM roughness index" M0:

$$\widehat{M0} = (M_i - \alpha_i) / \beta_i$$

The corresponding serviceability index is then

$$SIV = 5e^{-\left[\frac{\ln(32 M0)}{\beta_0}\right]^{\alpha_0}} \quad (4.3)$$

The parameters α_0 and β_0 will have been derived independently of any particular MRM and thus behave as constants. The new program prints a three-way table relating M_i , $\widehat{M0}$, and SIV. Other output includes confidence intervals for the parameters α_i and β_i , regression statistics SE and R^2 , and a plot of M_i vs $\widehat{M0}$.

Provisional Method

The new MRM standard $M0$ was devised in early 1978. To test its effectiveness without changing existing procedures a relationship was established between it and SI2 by "calibrating" it via the old method, as if it were an actual Maysmeter. This resulted in the serviceability index,

$$SIV = 5e^{-\left[\frac{\ln(32 M0)}{8.4933}\right]^{9.3566}} \quad (4.4)$$

Hence, the provisional method is simply to let the profile-derived SIV play the role of SI2 in the old method. While it is a definite improvement with respect to bringing different MRM units into alignment (standard errors of estimate are less than half as great), it should be replaced as soon as possible by the more direct linear method employing the revised MRMAL2 (Appendix C).

SUMMARY OF RESULTS FOR THE AUSTIN TEST SECTIONS

For the year beginning January 1978, both SI2 and SIV were recorded for the quarterly profilometer runs on the Austin test sections. The results are listed in Tables 4.1 and 4.2. Each measurement is actually the mean of two separate runs, with differences between repeats being at least three times larger for SI2 than for SIV. The SI2 results were also more erratic over the time period. Seventy percent of the variation of SIV about the means of 19 relatively unaltered sections could be explained by linear declines whereas only 42 percent of SI2 variation could be so explained. These facts suggest that the simpler statistic SIV, while correlating much better with Maysmeters, is also less sensitive to the measuring system and factors not related to the actual profile. Interestingly, the mean decline for these 19 sections over one year was only 0.22 serviceability units, which is comparable to the discrepancy between repeated SI2 measurements with the profilometer.

BENEFITS OF THE RECOMMENDED METHOD

In summary, the advantages of adopting the procedures described in Chapter 3, which employ statistic MO instead of SI as the Maysmeter calibration standard, are as follows:

(1) More accurate calibrations. This means that it will be safer to assume that the pavement is rougher than another when the determination is based on Maysmeter measurements made at different times or with different

TABLE 4.1. SERVICEABILITY INDICES FOR THE AUSTIN TEST SECTIONS
(SI2)

SEC	JAN 78	APR 78	JUL 78	OCT 78	JAN 79	APR 79
1	3.25	2.12	3.20	3.12	2.92	3.01
2	1.96	2.07	1.55	1.79	2.00	1.25
3	3.77	3.63	3.72	3.66	3.56	3.69
5	3.23	3.19	3.14	4.46	4.41	4.24
6	3.20	3.23	2.60	2.83	2.98	----
7	3.80	4.26	3.77	3.74	4.71	4.50
8	3.34	3.37	3.43	3.33	3.30	3.92
9	3.88	3.69	3.54	3.51	3.52	3.49
10	4.40	4.44	4.43	3.93	4.06	4.49
11	4.21	4.40	4.16	4.25	4.27	4.57
12	3.14	3.13	3.17	2.98	2.17	3.14
13	3.22	3.23	2.77	3.09	3.12	3.11
14	3.63	3.45	3.47	3.49	3.53	3.24
15	3.67	3.75	3.65	3.62	3.56	4.08
19	3.56	3.47	3.35	3.43	3.26	3.44
23	3.51	3.43	3.25	3.07	2.88	3.21
28	3.81	3.67	3.53	3.75	3.64	4.11
32	3.61	3.62	2.53	3.42	3.47	3.68
33	2.61	2.81	2.73	2.45	2.63	2.49
34	3.20	3.35	2.81	2.76	2.65	2.38
35	2.58	2.54	1.77	1.43	1.58	1.96
36	4.36	4.13	4.47	4.06	4.34	4.27
37	3.53	3.37	3.60	2.53	2.60	3.05
38	3.39	3.26	3.00	2.68	2.86	3.39
39	1.88	1.73	1.51	-.80	-.46	----
40	3.70	3.64	----	3.56	3.65	3.78
41	2.63	2.57	2.63	2.30	2.38	1.97
42	1.02	.96	.59	.79	.61	.05
43	1.77	2.04	1.65	2.14	1.92	1.49
44	2.40	2.25	2.62	2.15	2.21	2.00
45	.03	.39	-.06	.61	.10	-.65
SE:	.128	.183	.173	.121	.150	.161

SE TOTAL: .154 (181 Degrees of Freedom)

TABLE 4.2. SERVICEABILITY INDICES FOR THE AUSTIN TEST SECTIONS
(SIV)

SEC	JAN 78	APR 78	JUL 78	OCT 78	JAN 79	APR 79
1	3.37	3.23	3.09	3.17	3.15	3.02
2	1.84	1.86	1.62	1.61	1.67	1.33
3	3.67	3.64	3.53	3.75	3.76	3.88
5	3.45	3.38	3.29	4.70	4.66	4.68
6	2.70	2.68	2.31	2.54	2.40	----
7	4.54	4.58	4.24	4.48	4.83	4.81
8	3.78	3.68	3.71	3.69	3.68	3.66
9	3.97	3.97	4.02	3.99	4.01	3.19
10	4.65	4.61	4.67	4.59	4.59	4.50
11	4.34	4.19	4.14	4.11	4.13	4.48
12	3.15	3.09	3.09	2.90	2.11	2.19
13	2.81	2.83	2.72	2.63	2.65	2.47
14	3.86	3.75	3.74	3.73	3.73	3.37
15	3.27	3.41	3.36	3.36	3.38	3.51
19	3.59	3.47	3.48	3.44	3.19	3.75
23	2.97	2.92	2.62	2.69	2.54	2.64
28	4.14	4.09	3.77	4.02	3.98	4.13
32	3.58	3.58	2.91	3.45	3.51	3.51
33	2.78	2.77	2.73	2.47	2.41	1.56
34	2.99	2.98	2.79	2.68	2.68	2.33
35	1.45	1.53	1.75	1.62	1.59	1.23
36	4.15	4.24	4.41	4.17	4.21	4.31
37	3.23	3.07	2.91	2.71	2.32	2.50
38	2.47	2.53	2.41	2.19	2.20	2.03
39	1.88	1.65	2.01	1.11	1.16	----
40	3.32	3.39	----	3.48	3.56	3.53
41	2.29	2.25	2.31	2.03	2.19	1.93
42	1.07	1.10	.94	.82	.85	.81
43	1.25	1.24	1.15	1.91	1.99	1.70
44	1.40	1.38	1.34	1.21	1.23	1.13
45	1.01	1.02	.87	.67	.67	.41
SE:	.034	.062	.044	.034	.044	.050

SE TOTAL: .045 (177 Degrees of Freedom)

units. This is a benefit of the high correlation between Maysmeters and the profile statistic developed to simulate them.

(2) Simplicity. The calibration curves are much easier to deal with if they are linear. In this case the unit's calibration parameters, α_i and β_i , are the easily comprehended slope and intercept. Furthermore, the linear calibration model is by far the most widely used and studied; techniques for obtaining confidence intervals for parameters and estimates, etc., are well established.

(3) Adaptability to future requirements and conditions. The quality of SIV as an estimate of PSR (the mean of a subjective panel rating) might be expected to be less than that of SI2 because of the elimination of components not measurable by the Maysmeter; however, there are reasons to believe that the Maysmeter-derived SIV is not significantly worse than SI2 in predicting PSR and certainly much better for comparing the roughness of roads. Because the SI prediction equation (4.3) is just a rescaling of the calibrated unit's adjusted output, M_0 , and is free of the calibration process itself (i.e., parameters α_i and β_i), it is easily replaced when better PSR data are obtained. Important to the development of new and better relationships is the preservation of information provided by accurately calibrated Maysmeters.

CONCLUDING REMARKS

It is important not to confuse the problem of calibrating a group of instruments with the problem of interpreting their measurements. When the Texas Maysmeter calibration method was first devised, the Serviceability Index (SI) was the best available estimate of Present Serviceability Rating

Serviceability Rating (PSR), a measure of roughness which is meaningful. Since serviceability estimates were wanted from the Maysmeters, SI was chosen as the standard against which all units were to be calibrated. This would have been a good approach, however, only if Maysmeters were capable of measuring SI with as much accuracy as their precision would seem to indicate. Unfortunately, this is not the case. At best, Maysmeters can be assigned scalings so that different units give comparable "Maysmeter roughness" ratings. How the ratings should be used to predict other things, such as ride quality, is a problem to be considered apart from the calibration process itself.

To help clarify this point with an analogy, suppose that the readings of several homemade thermometers inserted in lakes of a given region correlated fairly well with the number of fish caught during the day. It would be desirable to know that one lake is a better fishing prospect than another, even though they were measured with different thermometers. A decidedly inferior approach to calibrating the thermometers would be to derive for each one of them, separately, a prediction equation by comparing its readings to the number of fish actually caught in a representative sample of lakes. Since the number of fish caught is only partially dependent on lake temperature we must expect that the individual equations derived from such a calibration procedure would be highly variable, depending on our sample lakes, the time of year, etc. Obviously, a much better approach would be to use a standard thermometer to correlate each homemade device with temperature, i.e., with something it is capable of measuring precisely. Then, with the benefit of results from all of the calibrated instruments, one

could seek a relationship between temperature and number of fish caught, number of fishermen, or whatever.

The analogy between Maysmeters and thermometers is not perfect, for it is not at all obvious what the equivalent of temperature in pavements should be. Our study of the Texas Maysmeters suggests, however, that a simple profile statistic based on RMSVA can serve effectively as a calibration standard. When the statistic is rescaled by regression techniques to approximate a serviceability rating, we find that different Maysmeters that are calibrated against it can measure roads and agree to within one or two tenths of a serviceability unit. This precision, of course, says nothing about the accuracy of such measurements as predictors of subjective serviceability ratings since the Maysmeter, like the thermometer, is necessarily limited in its response. However, quite apart from providing imperfect estimates of serviceability, it is evident that the Maysmeter is capable of measuring a certain kind of roughness with good precision. The obvious benefit of this is in making comparisons—for example, revealing differences in separate pavements and showing trends in deterioration or the effects of rehabilitation on roughness. It is for this purpose, especially, that a good calibration method based on a stable and valid reference is necessary.

REFERENCES

1. Walker, Roger S., and W. Ronald Hudson, "Analog-To-Digital System," Research Report 73-4, Center for Highway Research, The University of Texas at Austin, April 1970.
2. Walker, Roger S., and W. Ronald Hudson, "The Use of Spectral Estimates for Pavement Characterization," Research Report 156-2, Center for Highway Research, The University of Texas at Austin, August 1973.
3. Bergland, G. D., "A Guided Tour of the Fast Fourier Transform," IEEE Spectrum, July 1969.
4. Blackman, R. B., and J.W. Tukey, The Measurement of Power Spectra, Tukey, Dover Publications, Inc., New York, 1958.
5. Williamson, Hugh J., W. Ronald Hudson, and C. Dale Zinn, "A Study of the Relationships Between Various Classes of Road-Surface Roughness and Human Ratings of Riding Quality," Research Report 156-5F, Center for Highway Research, The University of Texas at Austin, August 1975.
6. Al-Rashid, Nasser I., Clyde E. Lee, and William P. Dawkins, "A Theoretical and Experimental Study of Dynamic Highway Loading," Research Report 108-1F, Center for Highway Research, The University of Texas at Austin, May 1972.
7. "Quarter Car Simulator Instruction Manual," K. J. Law Engineers, Inc., (Unpublished).
8. Hudson, W. Ronald, David McKenzie, and Grover Cunningham, "A Generalized Roughness Index for the Compatible Evaluation of Roads," Texas Research and Development Foundation, December 1980.
9. McKenzie, David, and Maitree Srinarawat, "Root Mean Square Vertical Acceleration (RMSVA) as a Basis for Mays Meter Calibration," Tech Memo BR-23 (unpublished), Center for Highway Research, The University of Texas at Austin, February 1978.
10. Roberts, Freddy L., and W. Ronald Hudson, "Pavement Serviceability Equations Using the Surface Dynamics Profilometer," Research

Report 73-3, Center for Highway Research, The University of Texas at Austin, April 1970.

11. Walker, Robert S., and W. Ronald Hudson, "A Correlation Study of the Mays Road Meter with the Surface Dynamics Profilometer," Research Report 156-1, Center for Highway Research, The University of Texas at Austin, February 1973.
12. Hu, Yi Chin, Hugh J. Williamson, and B. Frank McCullough, "A Study of the Mays Ride Meter," Research Report 177-3, Center for Highway Research, The University of Texas at Austin, January 1977.
13. Walker, Roger S., and W. Ronald Hudson, "The Use of Spectral Estimates for Pavement Characterization," Research Report 156-2, Center for Highway Research, The University of Texas at Austin, August 1973.
14. Williamson, Hugh J., and W. Ronald Hudson, "A Study of the Relationships Between Various Classes of Road-Surface Roughness and Human Ratings of Riding Quality," Research Report 156-5F, Center for Highway Research, The University of Texas at Austin, August 1975.
15. Scheffe, "A Statistical Theory of Calibration," The Annals of Statistics, 1973, Vol 1, p 1-47.

APPENDIX A

SUMMARY OF AVAILABLE PROGRAMS FOR
ROAD PROFILE ANALYSIS

PROGRAMS FOR PROFILE ANALYSIS

CONTENTS

Format for Digital Profilometer Tapes

ONESTEP

SI2

POWER

ROKYRD

DYMOL

QSIM

VERTAC

PLTPROF

Summary Table

PROGRAMS FOR PROFILE ANALYSIS

FORMAT FOR DIGITAL PROFILOMETER TAPES

Each of the computer programs described below accepts as input a magnetic tape which is already two stages of processing removed from the 4-track analog tape produced on board the profilometer. The analog data first undergo A-D processing on an HP2115 system. The resulting 9-track digital tape is then converted to a 7-track format which is compatible with The University of Texas CDC 6600 facility.

The 7-track tape is partitioned into two sets of files. Each file in the first set contains the data for one profilometer run on a section of interest, while files in the second set represent calibration runs in which voltages corresponding to a one-inch step in elevation were imposed. The program, ONESTEP, will accept a one-inch step file and compute the scale factors necessary to transform the data for one or more runs on a section to inches.

A data file consists of a leading two-word IDENT record, the two 60-bit words being used to store the file number, the number of records in the file, the road section number, and the run number. There follows a sequence of 300-word records containing the unscaled profile elevations packed as 12-bit (2's complement) binary integers. The digitization rate normally used (350 Hz) provides 5.92 data points for each foot of road length, with the right and left wheelpath data alternating. Hence, each tape record consists of $300 \times 60 \div 12 \div 2 = 750$ pairs of points representing $750 \div 5.92 = 126.7$ feet of profile.

The programs SI2, ROKYRD, VERTAC, QSIM, and PREDYML each have essentially the same file positioning and data unpacking routines, which can, if necessary, be replaced. The basic requirement of these programs is an array of alternating wheelpath data points and the scale factors necessary to convert these values to inches.

The above procedures for pre-processing profile data are to be revised when the Digital Profilometer replaces the original profilometer hardware because the tape formats, sampling rates, etc. will be different. Minor revisions of the analysis programs will also be required.

ONESTEP

This program accepts the one-inch step files on a digital profilometer tape and computes the right and left wheelpath scale factors. During processing by the other programs, each point (integer) on a data file is converted to inches of elevation by dividing by the appropriate scale factor (typically, a number between 100 and 1000). Hence, the conversion to actual elevations is delayed until the very last stage of processing (Ref 1).

SI2

Used routinely since 1974 to obtain serviceability estimates from a road profile, this program employs the Fast Fourier Transform (FFT) algorithm to extract a set of power spectral averages. Regression on these averages (based on a 1968 panel rating of 86 Texas roads) yields a serviceability

estimate (SI) for each file (run) processed. The user specifies the scale factors (obtained from ONESTEP), the file number, and whether the pavement is rigid or flexible. The program simply outputs this information along with the single statistic, SI.

The regression equation incorporated in SI2 is based on a model whose independent variables are mean wavelength amplitudes from 64 discrete wavelength bands. There are 22 terms in the equation, including a constant and a dummy variable (flexible or rigid). The variables correspond to wavelengths ranging from 8.6 feet to 86 feet. The smaller wavelengths were excluded, not because they were poorly correlated with PSR but because of excessive tape recorder noise on the analog tapes from the rated sections. The model, however, was sufficient to explain 89 percent of the variation in panel ratings with a standard error of residuals of .33.

A limitation of SI2 is its lack of generality. It requires elevation data at $1/5.92 = 0.169$ -ft. intervals expressed as 6.750 right-left wheelpath sample pairs records to define one section 1,140 feet long. The program is not configured so that it can be easily revised without changing to some extent the nature of its serviceability estimate. The measured data points are first reduced in number by 75 percent through application of a linear group-smoothing technique. After removing trends that can distort the low frequency spectral estimates, a "cosine taper-window" is used to correct for end effects—a process that effectively gives less weight to profile characteristics in the first and final 114 feet of the section length. As the spectral values produced by the FFT algorithm are themselves statistical estimates, it was considered appropriate to base them on a fixed quantity of data (Ref 2).

POWER (VERSION 6.0)

Used primarily as a research tool, this program had an important role in the development of the spectral analysis methods for evaluating pavement roughness. POWER utilizes the FFT algorithm to compute a set of power spectral estimates for a specified number of bands. If this number is 64, then a routine is called to compute the same serviceability estimate that the more recent program, SI2, provides. The user may also request printer plots of such statistics as slope variance, coherence, power, and log power.

The computational techniques include low pass filtering to avoid any distortion by subsampling, detrending (subtracting from each series its least squares regression line), and windowing to reduce the effects of finite sample size on the Fourier Transform. These methods are described in references 4 and 5.

Unlike program SI2, POWER permits its user to specify both a decimation rate and the number of points to be used in the computations. Thus, a section of desired length may be analyzed provided the corresponding number of points (after decimation) in each wheelpath is at least four times the number of frequency bands and not more than 2048.

The program provides a summary table listing the band frequencies (cycles per foot), power spectral estimates for right and left profiles (P1 and P2), and the cross-power estimates along with their coherence and phase. A more detailed description of this output can be found in Appendix 1 of reference 4.

ROKYRD (VERSION 2.3)

This program employs digital filtering techniques to obtain a collection of statistics to characterize the measured road profile. The statistics are then used to derive serviceability estimates corresponding to different types of roughness, with separate relationships assumed for flexible and rigid pavements.

The computational method consists of three parts. First, the differences between the outputs of successive low-pass filters are used to isolate irregularities in contiguous passbands. The user can specify the band limits; however, the serviceability equations require data for wavelengths of 4 to 10 feet, 10 to 25 feet, 25 to 50 feet, and 50 to 100 feet. Next, for each passband, a moving root-mean-square amplitude is computed at each point in the corresponding artificial (filtered) profile. Values corresponding to the first and final portions of the section--usually, a distance equal to the longest specified wavelength--are excluded because of distortion by end effects. The final step is to compute the sample frequency distribution of these mean amplitudes.

ROKYRD provides a summary table which lists the mean, standard deviation, and percentiles 50, 75, 90, 95, and 99 for the amplitudes (in inches) for each of the requested passbands. This is done for each wheelpath and for the difference profile: right wheelpath minus left wheelpath. Also, the user of the program may request Calcomp plots of the profile trace of either wheelpath (or of the difference). Both filtered and unfiltered profiles can be plotted, either separately or in the same frame.

Although the serviceability equations built into ROKYRD are derived from essentially the same data (i.e., the 1968 Texas panel rating), they differ in several important respects from the one in program SI2:

(1) Instead of power spectral means, the 50th and 90th percentile values of filtered profile amplitudes are used as independent variables. Thus, the model can in principle distinguish between roads whose roughness is either highly localized or relatively uniform.

(2) Separate models are incorporated to handle the two types of roads: flexible and rigid. The equation for flexible pavements contains eight terms and was sufficient to explain 83 percent ($R^2 = .83$) of the mean rating variability with a residual standard error (SE) of .38. Similar results were obtained for the rigid pavement model ($R^2 = .83$, SE = .32), which contains only six terms, including a dummy variable intended to explain the "auditory or visual differences" between JRCP and CRCP.

(3) Development of these models, like that of SI2, was hindered by the lack of good data in the higher frequency range; however, in this case, an initial filtering operation was believed to have been at least partially successful in eliminating distortion in the 4 to 10-ft wavelength area. Thus, the terms in the ROKYRD equations encompass a slightly broader range of profile wavelengths.

In addition to the overall serviceability estimate, combined, transverse, and longitudinal SI values for each of the four passbands are included in the summary table. Of interest is the fact that the models based solely on wavelengths of 4 to 10 ft fit the original data almost as well as the complete models: $R^2 = .74$ and SE = .45 for flexible pavements and $R^2 = .74$ and SE = .37 for rigid pavements.

ROKYRD is a general program in that it allows the user to specify the step size of digitized data (usually .169 ft), the rate at which these data are to be subsampled, and the wavelength limits of up to 10 passbands to be analyzed. Unlike SI2, a road section of arbitrary length is acceptable provided it is at least more than twice that of the longest profile wavelength of interest (Ref 6).

DYMOL

DYMOL is a generalized mathematical model which characterizes the dynamic behavior of five different classes of highway vehicles. The model consists of a series of interconnected masses, springs, and dashpots and is used to predict the magnitude, duration, and location of dynamic wheel loads applied normally to the roadway surface by the wheels of single unit and articulated vehicles operating under various conditions. The model may be forced upon a natural road profile which is recorded in the field by a profilometer and converted to a suitable digital format.

The sets of differential equations used to describe the motion and calculate the forces between the tire and the road surface are solved by a numerical technique that is based on the assumption that the displacement, the velocities, and the accelerations of the system are known at any particular time. The time increment h may be taken as approximately $1/5$ to $1/6$ of the shortest period of oscillation, t , to assure convergence and stability of the solution. For the purpose of solving the sets of equations, the value for the wheel to travel a distance of one inch or 0.001 second, whichever is greater, is normally used. This criterion keeps the h -value

within the range for convergence and stability. The model is limited to speeds between 10 and 60 mph (176 to 1056 ips).

A tire envelope of 8 inches is estimated by taking a moving average of 46 points (steps) at 0.176 inch, with the average value of these points representing the center of the envelope. DYMOL calculates this as the profile input to the differential equations of motion. The enveloping process allows the model to more accurately represent the tire action of absorbing the effects of small bumps or depressions without displacing the axle.

The program could be modified to accept profile data at larger step sizes. However, to get a reasonable moving average for the tire envelope, at least five points should be used, which implies a maximum step size of two inches.

A separate program, PREDYML, functions as a preprocessor for DYMOL. It accepts a digitized (4000 Hz for 0.176-inch step) profilometer tape along with calibration factors from ONESTEP, unpacks the data, scales the data, and writes onto a new tape only the portion of profile the DYMOL is to be run on in its unpacked and scaled form.

The output of DYMOL consists of printed tables and/or Calcomp plots of the dynamic wheel forces corresponding to distance down the roadway. The printed tables also show the excitations to the tires which are derived from the smoothed profile data. The plots, which are provided only for the left wheel, can portray either the dynamic wheel forces or the axle displacements (Ref 7).

QSIM (VERSION 1.0)

This program is basically a digital version of the Quarter Car Simulator, an electronic device developed by K. J. Law, Inc., to extract from the road profile a roughness measure of a specific type. The Quarter Car Index (QI) was designed to replicate, by means of a mathematical model consisting of a tire-wheel mass, suspension spring, shock absorber, and vehicle mass, the Roughness Index (RI) of the Bureau of Public Roads Roughometer.

In QSIM, the differential equations describing the motion of the model components (i.e., the sprung and unsprung masses) are solved by digital, rather than analog, methods. More specifically, the transition matrix method is applied to the series of profile elevations from a digitized profilometer tape. The relative motion, in the upward direction, between the wheel axle and the sprung frame is integrated for a fixed distance along the road. The result is QI, which has units of inches per mile.

In addition to the tape input, the user of QSIM provides information such as step-size of digitized data, scale factors, section identification method, and speed at which the quarter-car is to be simulated. A preliminary comparison between the electronic (analog) and digital versions of QI on several Brazilian roads showed very good agreement between the methods (Refs 8 and 9).

VERTAC

Developed as part of research done in early 1978 to derive a Maysmeter calibration standard, VERTAC provides a set of roughness statistics closely related to the physical concept of vertical acceleration. The basic algorithm, which is numerically much simpler than either spectral analysis or digital filtering, can be stated here.

Let Y_1, Y_2, \dots, Y_n be elevations in inches or equally spaced points along one wheelpath of the profile and let b be the distance between adjacent points. Then a simple estimate of the second derivation of Y at point i is

$$(S_b)_i = (Y_{i+1} - 2Y_i + Y_{i-1})/b^2$$

Hence, the root-mean-square vertical acceleration (RMSVA) at base length b is defined as

$$VA_b = C \left(\frac{\sum_{i=2}^{N-1} (S_b)_i^2}{(N-2)} \right)^{\frac{1}{2}}$$

where C is the constant that transforms units to ft/sec for a given vehicle speed. By default, a 50 mph speed is assumed.

An examination of profilometer data will reveal that the index VA increases sharply as b decreases, but that VA is selectively sensitive to roughness amplitudes of wavelengths $2b$. VERTAC lists the RMSVA indices at base lengths $b = .5, 1, 2, 4, 8, 16, 32,$ and 65 feet for each wheelpath.

Also provided is a Maysmeter roughness estimate in counts/.2 mile:

$$Mo = -20 + 23VA_4 + 58VA_{16} ,$$

and a serviceability estimate based on Maysmeter roughness:

$$SIV = 5 \exp \left[- \left(\frac{\ln(32 MO)}{8.49331} \right)^{9.35661} \right]$$

The index Mo was adequate to describe quite well the response of eight Texas Maysmeters on 29 road sections near Austin. The measurement of a particular Maysmeter, for example, is predicted by $M_k = A_k + B_k Mo$, where A_k and B_k are calibration constants close to 0 and 1 (trailer units) or to 10 and 1.5 (car units). For a typical Maysmeter trailer, the standard error of prediction is less than 10 counts/.2 mile while the mean roughness of the Austin test sections is about 100 counts/.2 mile.

The index SIV was obtained by calibrating the simulated Maysmeter Mo against the serviceability index computed by program SI2--the same technique used in the past for calibrating the Texas Maysmeters. In effect, the RMSVA index is scaled to a number between 0 and 5.

In order to identify spurious data on the tape, VERTAC writes on a separate output file the maximum elevations between adjacent points on each data record and identifies all points with such differences greater than one inch. This was found to be necessary because of occasional faults in the digitization process.

PLTPROF

PLTPROF is a program that takes a digitized profilometer tape and plots the profile on a Calcomp plotter using ONESTEP calibration factors and an input step size corresponding to the digitization rate. It can plot any portion of the profile specified on any X and Y scale within limits of the plotter (.005 in. is the smallest increment).

SUMMARY OF ROAD PROFILE ANALYSIS PROGRAMS

Name	Description	User Input	Profile Input	Output	Lines of Code
ONESTEP	Computes wheel path scale factors from one-inch step files	First file no., no. of files to process	Tape	List	178
SI2	Computes present serviceability estimate from spectral estimates	Section ID, scale factors, file no., pavement type	Tape	List	536
POWER 6	Performs power spectral analysis of road profile	No. points, scale factors, no. bands, decimation rate, plot options	Tape	Tables/ printer plots	722
ROKYRD	Provides amplitude statistics and serviceability estimates from filtered profiles	File no., section limits, step size, decimation rate, band options, plot options, pavement type	Tape or cards	Tables/ plots	1007
DYMOL	Models the dynamic wheel forces of five classes of trucks	Truck class, speed, plot/ print options	Tape from PREDYML	Tables/ plots	3084
PREDYML	Writes an unpacked and scales data tape for DYMOL	File no., scale factors, step size, section limits	Tape	Tape for DYMOL	82
QSIM	Computes a Quarter Car Index from road profile	Section ID, scale factors, step size, file no., speed of simulated vehicle	Tape or cards	List	410
VERTAC	Computes vertical acceleration (RMSA) indices and simulates Mays Meter roughness	Section ID, scale factors, step size, file no., speed of simulated vehicle	Tape or cards	Tables	
PLTPROF	Provides Calcomp plots of unfiltered profile	File no., axes lengths, step size, section limits, wheelpath, scale factor	Tape	Plots	342

APPENDIX B

VERTAC--USER'S GUIDE
AND SOURCE LISTING

VERTAC--USER'S GUIDE AND SOURCE LISTING

DESCRIPTION OF ITEMS ON THE VERTAC PRINTOUT

1. Title (usually a road name), date of profilometer run, section number, and run number.
2. Length of section actually processed (in feet), along with profilometer tape format and positioning information.
3. GAINR, GAINL - right wheelpath and left wheelpath scale factors used by program to convert profile elevations to proper units.
4. A row of roughness indices corresponding to each RMSVA baselength examined - normally, multiples of the sampling interval (.169 ft) in the range .5 ft to 130 ft:

Col 1 base length (feet)

Col 2 right wheelpath RMSVA

Col 3 left wheelpath RMSVA

Col 4 mean RMSVA from both wheelpaths

Col 5 SI value - Eq 2 with M_0 replaced with a linear function
the mean RMSVA (col 4)

5. MRM Roughness (counts/.2 mile) - equivalent to MO (Eq 1) with units inches/mile.

6. Flexible Pavement Serviceability - SIV (Eq 2).

```

PROGRAM VERTAC/INPUT=65,DATA=513,OUTSAV=65,OUTPUT=65,
TAPF1,TAPF2=DATA,TAPF3=OUTSAV,TAPF5=INPUT,TAPF6=OUTPUT)
VE 1
VE 2
VE 3
C*****
VE 4
C
VE 5
C
VE 6
C
VE 7
C
VE 8
C
VE 9
C
VE 10
C
VE 11
C
VE 12
C
VE 13
C
VE 14
C
VE 15
C
VE 16
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VE 17
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VE 18
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VE 19
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VE 21
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VE 26
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VE 27
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VE 28
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VE 29
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VE 30
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VE 31
C
VE 32
C
VE 33
C*****
VE 34
C
VE 35
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VE 36
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VE 37
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VE 38
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VE 39
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VE 40
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VE 41
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VE 42
C
VE 43
C
VE 44
C
VE 45
C
VE 46
C
VE 47
C
VE 48
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VE 49
C
VE 50
C
VE 51
C
VE 52
C
VE 53
C
VE 54
C
VE 55

```

THIS PROGRAM COMPUTES THE ROAD PROFILE STATISTIC ROOT-MEAN-SQUARE VERTICAL ACCELERATION (RMSVA) AT A SEQUENCE OF USER-SPECIFIED BASE LENGTHS (MULTIPLES OF DATA SAMPLING INTERVAL). THE HAYS METER SIMULATION MR (MRM) IS COMPUTED PROVIDED THE COMPONENT RMSVA INDICES (FOR BASE LENGTHS 4 FEET AND 1A FEET) ARE ALSO COMPUTED. IN ADDITION, THE SERVICEABILITY INDEX STV, WHICH IS A FUNCTION OF MR, IS OUTPUT FOR EACH MEASURED ROAD SECTION. DATA SMOOTHING AND SUBSAMPLING OPTIONS ARE PROVIDED FOR THE USER.

WRITTEN BY
 DAVID MCKENZIE
 PRENTISS RIDDLE
 CENTER FOR TRANSPORTATION RESEARCH
 THE UNIVERSITY OF TEXAS AT AUSTIN
 AUSTIN, TEXAS 78712

INPUT FORMAT

NOTES: CARDS 1 AND 2 FORM A PAIR WHICH MAY BE REPEATED FOR AS MANY RUNS AS DESIRED.

CARD 0 -- FORMAT(A6)

IDATE DATE

CARD 1 -- FORMAT(8A10)

TI RUN INFORMATION

CARD 2 -- FORMAT(4I5,4F10.0,2F5.0,2A11)

IFILE NO. OF FILE ON TAPF1. IF LE 0, IFILE WILL BE SET EQUAL TO VE
 ONE. FILE NOS. FOR SUCCESSIVE SECTIONS MUST BE IN
 ASCENDING ORDER.

NREC NO. OF FIRST RECORD ON FILE TO BE READ. IF NREC LT 0,


```

C*****VE 118
C COMMON /SMTH/ TSM,C(21),SUMC(2),IDEC,TC,NXOR,NXDA,NPR VE 119
COMMON /PACK/ TD(5),IN(300) VE 120
COMMON /STAT/ NOCHK,RESS,RMAX,MAXR,ORSR(21),LIMR,SCL(2) VE 121
COMMON // DAT(2,7000),ORS(1500) VE 122
C DIMENSION IO(1500), TI(8), V(2), RM(3,9), STEP(9), TODAY(2,1) VE 123
DIMENSION RL(9), AR(9), A1(9), SIVA(9) VE 124
EQUIVALENCE (IO,ORS), (DAT,IODAT) VE 125
LOGICAL NOCHK VE 126
C DATA NEQ/9/ VE 127
DATA BL /0.5, 1.0, 2.0, 4.0, 8.0, 16.0, 32.0, 64.0, 128.0/ VE 128
DATA AR /52.715, 16.162, -28.588, -23.506, 6.125, 10.831, 10.103, VE 129
19.275, 26.483/ VE 130
DATA A1 /0.56721, 2.9354, 13.383, 34.864, 66.132, 139.1A, 296.66, VE 131
602.00, 1603.8/ VE 132
DATA LREC/0/,MAXN/7000/,MAXSTP/9/,SPEED/50./,MXISM/10/ VE 133
DATA B0,B1,B2/-20.,23.,58./,DLIM/1.0/ VE 134
C REWIND 2 VE 135
REWIND 4 VE 136
READ (5,10) IDATE VE 137
WRITE (4,10) IDATE VE 138
10 FORMAT (A6) VE 139
C READ TITLE CARD AND DATA CHARACTERISTICS VE 140
C NPG = 1 VE 141
NRN = 0 VE 142
20 READ 30, TI VE 143
30 FORMAT (A10) VE 144
IF (FOP,5) 40,50 VE 145
40 CALL EXIT VE 146
50 READ 60, IFILE,NREC,IDEC,ISM,SCL,DIST,SKP,STEPIN,RASE1,NWR,NCHK VE 147
60 FORMAT (4I5,4F10.0,2F5.0,2A1) VE 148
IF (NRN.EQ.0) PRINT 70, NPG VE 149
70 FORMAT (1H1/1H0,6X,31HPROGRAM VERTAC = VERSION 2.1 = ; VE 150
16HREVISED 00/21/81,12X,4HPAGE,13//1H0,6X,66(1H*)) VE 151
PRINT 80, TI VE 152
80 FORMAT (1H0,6X,8A10) VE 153
IF (IFILE,LE,0) IFILE = 1 VE 154
IF (RASE1,LE,0) RASE1 = .55 VE 155
IF (STEPIN,LE,0) STEPIN = 0.169 VE 156
IF (DIST,LE,0) DIST = 1050. VE 157
IF (IDEC,LE,0) IDEC = 1 VE 158
IF (NREC.EQ.0) NREC = 1 VE 159
INF = 1 VE 160
IF (NREC,LI,0) INF = IFILE = 2 VE 161
NOCHK = 'F' VE 162
IF (NCHK.EQ.1HN .OR. ISM.EQ.0) NOCHK = 'T' VE 163
C INITIALIZE FOR NEW SECTION VE 164
C NRN = NRN+1 VE 165
MDIST = DIST+.5 VE 166
NSKP = SKP/STEPIN VE 167
MSKP = NSKP*STEPIN+.5 VE 168
PRINT 90, MDIST,MSKP,NREC,IFILE VE 169
90 FORMAT (/6X,15, 18H FT SECTION BEGINS,14, 21H FT FROM START OF RECVE VE 170
.I3, 8H OF FILE,13) VE 171

```

	NP = DIST/STEPIN	VE 180
	N = NP+NSKP	VE 181
	N2 = N+N	VE 182
	NS = NRKP+NSKP	VE 183
	MX = (NP-2*ISM)/IDFC	VE 184
	IF (MX.GT.MAXN) GO TO 870	VE 185
C		VE 186
C	UNPACK DATA FROM TAPE1 OR FILE <DATA>	VE 187
		VE 188
	IF (INF.EQ.1) CALL POSITP (IFILE,NREC,LREC,NSEC)	VE 189
	NWORDS = 300	VE 190
	NXDA = 1	VE 191
	CALL INITSM (ISM,C)	VE 192
	NXDA = NSKP+ISM+1	VE 193
	RMAX = 0.	VE 194
	SUMC(2) = RMAX	VE 195
	SUMC(1) = SUMC(2)	VE 196
	RESS = SUMC(1)	VE 197
	IC = 0	VE 198
	NRPTS = N2	VE 199
100	NPR = MIN0(750,NRPTS/2)	VE 200
	L = 2*NPR	VE 201
	IF (INF.EQ.1) READ (1) IN	VE 202
	IF (INF.EQ.2) READ (2,110) (IO(K),K=1,L)	VE 203
110	FORMAT (14I5)	VE 204
	IF (EOF,INF) 120,140	VE 205
120	PRINT 130, N,IFILE	VE 206
130	FORMAT (6X, 47HEND OF FILE ENCOUNTERED WHEN ATTEMPTING TO READ,IS, 26H PAIRS OF POINTS FROM FILE,I3)	VE 207
	STOP 3	VE 208
140	IF (INF.EQ.2) GO TO 150	VE 209
	LRFC = LREC+J	VE 210
	IF (NRPTS.LT.1500) NWORDS = (NRPTS+4)/5	VE 211
	NXTIO = 1	VE 212
	CALL UNPACK (NWORDS,NXTIO,IO)	VE 213
150	DO 160 I=1,L	VE 214
160	ORS(I) = IO(I)	VE 215
	CALL SMOOTH (ORS,DAT)	VE 216
	NRPTS = NRPTS-1500	VE 217
	IF (NRPTS.GT.0) GO TO 100	VE 218
C		VE 219
	N = NXDA-1	VE 220
	N2 = 2*N	VE 221
	RESE = 0.	VE 222
	IF (.NOT.NOCHK) RESE = SQRT(RESS*2/N2)	VE 223
	IF (NWR.NE.1HN) GO TO 190	VE 224
		VE 225
C		VE 226
C	PLACE UNSCALED DATA ON TAPE2 = DATA.	VE 227
		VE 228
	REWIND 2	VE 229
	DO 170 K=1,N	VE 230
	DO 170 I=1,2	VE 231
170	IODAT(I,K) = DAT(I,K)+IS	VE 232
	WRITE (2,110) ((IODAT(I,K),I=1,2),K=1,N)	VE 233
	DO 180 K=1,N	VE 234
	DO 180 I=1,2	VE 235
180	DAT(I,K) = IODAT(I,K)	VE 236
		VE 237
C		VE 238
C	SCALE TO INCHES AND COMPUTE VERTICAL ACCELERATION.	VE 239
		VE 240
190	STEPIN = IDEC*STEPIN	VE 241
	IF (ISM.NE.0) PRINT 200, STEPIN,ISM,RESE	VE 242

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200 FORMAT (6X, 10H SAMPLING INTERVAL, F5.2, 5X, 4HISM, T, P, 5X, 3HSP1, VE 242
.F6.4) VE 243
PRINT 210, SCI VE 244
210 FORMAT (10H, 16X, 6HGAINR, F7.1, 12X, 6HGAINL, F7.1) VE 245
IF (SCL(1).LE.0. .OR. SCL(1).GE.2000.) GO TO 490 VE 246
IF (SCL(2).LE.0. .OR. SCL(2).GE.2000.) GO TO 490 VE 247
DO 220 I=1,N VE 248
DAT(1,I) = DAT(1,I)/SCL(1) VE 249
DAT(2,I) = DAT(2,I)/SCL(2) VE 250
220 CONTINUE VE 251
C VE 252
C WRITE OUT MAXIMUM DISPLACEMENTS VE 253
C VE 254
K = 1 VE 255
DIF1 = 0. VE 256
DO 230 I=2,N VE 257
DIF2 = AMAX1( DIF1, ABS(DAT(1,I)-DAT(1,I-1)), VE 258
ABS(DAT(2,I)-DAT(2,I-1)) ) VE 259
IF (DIF2.GT.DIF1) K = I VE 260
DIF1 = DIF2 VE 261
230 CONTINUE VE 262
PRINT 240, DIF1, K*(STEPIN/IDEC) VE 263
240 FORMAT (90X, 4H STEP, F6.2, 7H IN AT, F7.1, 4H FT.) VE 264
C VE 265
NSTEP = MAX1(BASE1/STEPIN, 1) VE 266
NMSTEP = 0 VE 267
C VE 268
C WRITE SCALED DATA IN FORMAT SUITABLE FOR PLOTTING VE 269
C VE 270
IF (NWR.NE.1HS) GO TO 270 VE 271
REWIND 2 VE 272
TIM = 0. VE 273
DO 250 I=1,N VE 274
WRITE (2,260) TIM, DAT(1,I), DAT(2,I) VE 275
250 TIM = TIM+STEPIN VE 276
260 FORMAT (3F10.4) VE 277
C VE 278
270 INC2 = NSTP+NSTP VE 279
NM = N+INC2 VE 280
IF (NM.LT.1) GO TO 290 VE 281
NMSTEP = NMSTEP+1 VE 282
V(2) = 0. VE 283
V(1) = V(2) VE 284
DO 280 I=1,2 VE 285
DO 280 J=1,NM VE 286
280 V(I) = V(I)+(DAT(I,J)-2.*DAT(I,J+NSTP)+DAT(I,J+INC2))*2 VE 287
STEP(NMSTEP) = NSTP*STEPIN VE 288
TIM = (12.*STEP(NMSTEP)/(17.6*SPEED))**2 VE 289
RM(1,NMSTEP) = SQRT(V(1)/NM)/(TIM*12.) VE 290
RM(2,NMSTEP) = SQRT(V(2)/NM)/(TIM*12.) VE 291
RM(3,NMSTEP) = (RM(1,NMSTEP)+RM(2,NMSTEP))*S VE 292
IF (NMSTEP.GE.MAXSTP) GO TO 295 VE 293
NSTP = 2*NSTP VE 294
GO TO 270 VE 295
C VE 296
290 MAXSTP = NMSTEP VE 297
IF (NMSTEP.LE.0) GO TO 510 VE 298
295 CONTINUE VE 299
C VE 300
C CALCULATE SI VALUES ASSOCIATED WITH VERTICAL ACCELERATIONS VE 301
C VE 302
DO 320 K=1,MAXSTP VE 303

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510 PRINT 520	VE 366
520 FORMAT (6X, 46H --- NOT ENOUGH POINTS TO COMPUTE FIRST RMSVA)	VE 367
STOP 3	VE 368
C	VE 369
END	VE 370

FUNCTION SIM0 (RMO)	SI 1
C	SI 2
C FUNCTION TO CONVERT AN MO VALUE TO AN SI VALUE.	SI 3
C	SI 4
C DATA ALPHA,BETA / 9.35661, 8.49331 /	SI 5
C	SI 6
C IF (RMO .LE. 0.0) GO TO 100	SI 7
C	SI 8
C SIM0 = 5.0*EXP(-(ALOG(32.0*RMO)/BETA)**ALPHA)	SI 9
C RETURN	SI 10
C	SI 11
100 SIM0 = 5.0	SI 12
C RETURN	SI 13
C	SI 14
END	SI 15

FUNCTION PCTERR (A , B)	PC 1
C	PC 2
C ROUTINE TO CALCULATE PERCENTAGE ERROR (ABS(A-B) AS A PERCENTAGE	PC 3
C OF ABS(B)).	PC 4
C	PC 5
C IF (R .EQ. 0.0) GO TO 100	PC 6
C	PC 7
C PCTERR = ABS((A-B)/B) * 100.0	PC 8
C RETURN	PC 9
C	PC 10
100 PCTERR = ABS(A)	PC 11
C RETURN	PC 12
C	PC 13
END	PC 13

SUBROUTINE INTYSM(TSM,C)	IN 1
C	IN 2
C COMPUTES SMOOTHING COEFFICIENTS FOR OBSERVATIONS OF	IN 3
C POLYNOMIAL TRENDS OF DEGREE THREE.	IN 4
C	IN 5
C DIMENSION C(21)	IN 6
C	IN 7
C C(1)=1	IN 8
C IF(I=1,0) RETURN	IN 9
C IF(I=GT,1) GO TO 20	IN 10
C DO 10 I=1,3	IN 11

10	C(I)=1/3; RETURN	IN	12
C		IN	13
C		IN	14
20	K1=9*(ISM+ISM+ISM)-3 S2=(2*ISM-1)*(2*ISM+1)*(2*ISM+3) DO 30 I=0,ISM	IN	15
		IN	16
		IN	17
30	C(ISM+1-I)=C(ISM+1+I)=(K1-15*I+I)/S2	IN	18
	RETURN	IN	19
	END	IN	20

	SUBROUTINE SMOOTH(OBS,DAT)	SM	1
	COMMON /SMTH/ISM,C(21),SUMC(2),IDEC,IC,NXOR,NXDA,NPR	SM	2
	COMMON /STAT/ NOCHK,RESS,RMAX,MAXR,OBRS(21),LIMR,SCL(2)	SM	3
	DIMENSION OBS(2,1),DAT(2,1),OBSL(2)	SM	4
	LOGICAL NOCHK	SM	5
	DATA OBRS/0,0/	SM	6
C		SM	7
	MAXC=ISM+ISM+1	SM	8
C		SM	9
5	M1=MAXR(1,NXOR-ISM)	SM	10
	IF(M1.GT.NPR) GO TO 10	SM	11
	M2=MIN0(NXOR+ISM,NPR)	SM	12
	IF(NOCHK.OR.NXOR.LT.M1.OR.NXOR.GT.M2) GO TO 6	SM	13
	OBRS(1)=OBS(1,NXOR)	SM	14
	OBRS(2)=OBS(2,NXOR)	SM	15
C		SM	16
6	DO 10 M=M1,M2	SM	17
	IC=IC+1	SM	18
	DO 10 I=1,2	SM	19
10	SUMC(I)=SUMC(I)+OBS(I,M)*C(IC)	SM	20
C		SM	21
	IF(IC.LT.MAXC) GO TO 30	SM	22
	IC=0	SM	23
	DO 20 I=1,2	SM	24
	DAT(I,NXDA)=SUMC(I)	SM	25
	IF(NOCHK) GO TO 20	SM	26
C		SM	27
	RES=(OBRS(1)-SUMC(1))/SCL(1)**2	SM	28
	RESS=RESS+RES	SM	29
	IF(RPS.LE.RMAX) GO TO 20	SM	30
	RMAX=RES	SM	31
	MAXR=NXDA	SM	32
	K=0	SM	33
	DO 15 M=M1,M2	SM	34
	K=K+1	SM	35
15	OBRS(K)=OBS(I,M)	SM	36
	LIMR=K	SM	37
C		SM	38
20	SUMC(I)=0.	SM	39
C		SM	40
	NXDA=NXDA+1	SM	41
	NXOR=NXOR+IDEC	SM	42
	GO TO 5	SM	43
C		SM	44
30	NXOR=NXOR-NPR	SM	45
	RETURN	SM	46
	END	SM	47

PROGRAM VERTAC - VERSION 2.2 - REVISED 04/21/81

PAGE 1

AUSTIN TEST --- APRIL 1981 --- SEC. 1, RUN 1

1050 FT SECTION BEGINS 0 FT FROM START OF REC 1 OF FILE 2
(FILE: 2 SEC: 1 RECORDS: 9)

GAINR: 283.4 GAINL: 278.1

RMS VERTICAL ACCELERATION (FT/SEC SQ) AT 50 MPH:

BASE LENGTH	RIGHT	LEFT	COMBINED	ESTIMATED ST
.5	78.18	77.29	77.74	2.75
1.0	26.26	26.19	26.23	2.82
2.0	8.02	8.01	8.02	3.13
4.1	2.81	2.63	2.72	3.32
8.1	1.17	.96	1.06	3.18
16.2	.56	.52	.54	2.98
32.4	.33	.30	.31	2.82
64.9	.15	.14	.14	2.59
129.8	.06	.04	.05	2.47

MRM ROUGHNESS (COUNTS/.2 MILE): 73.84

FLEXIBLE PAVEMENT SERVICEABILITY: 3.20

AUSTIN TEST --- APRIL 1981 --- SEC. 1, RUN 2

1050 FT SECTION BEGINS 0 FT FROM START OF REC 1 OF FILE 3
(FILE: 3 SEC: 1 RECORDS: 9)

GAINR: 283.4 GAINL: 278.1

RMS VERTICAL ACCELERATION (FT/SEC SQ) AT 50 MPH:

BASE LENGTH	RIGHT	LEFT	COMBINED	ESTIMATED ST
.5	74.04	76.00	75.02	2.78
1.0	25.36	26.21	25.79	2.85
2.0	8.12	7.95	8.04	3.13
4.1	2.89	2.65	2.77	3.28
8.1	1.27	.96	1.11	3.11
16.2	.60	.52	.56	2.91
32.4	.33	.32	.32	2.58
64.9	.14	.14	.14	2.57
129.8	.05	.05	.05	2.53

MRM ROUGHNESS (COUNTS/.2 MILE): 76.20

FLEXIBLE PAVEMENT SERVICEABILITY: 3.19

AUSTIN TEST --- APRIL 1981 --- SEC. 2, RUN 1

1050 FT SECTION BEGINS 0 FT FROM START OF REC 1 OF FILE 4
(FILE: 4 SEC: 2 RECORDS: 9)

GAINR: 378.2 GAINL: 374.1

RMS VERTICAL ACCELERATION (FT/SEC SQ) AT 50 MPH:

BASE LENGTH	RIGHT	LEFT	COMBINED	ESTIMATED SI
.5	102.14	86.37	94.25	2.57
1.0	40.66	30.44	35.75	2.31
2.0	15.62	10.90	13.26	1.89
4.1	7.17	4.54	5.86	1.53
8.1	3.04	1.85	2.45	1.63
16.2	1.19	.81	1.00	1.87
32.4	.48	.39	.44	2.03
64.9	.22	.18	.20	1.99
129.8	.07	.07	.07	1.90

MRM ROUGHNESS (COUNTS/.2 MILE): 172.74

FLEXIBLE PAVEMENT SERVICEABILITY: 1.50

AUSTIN TEST --- APRIL 1981 --- SEC. 2, RUN 2

1050 FT SECTION BEGINS 0 FT FROM START OF REC 1 OF FILE 5
(FILE: 5 SEC: 2 RECORDS: 9)

GAINR: 378.2 GAINL: 374.1

RMS VERTICAL ACCELERATION (FT/SEC SQ) AT 50 MPH:

BASE LENGTH	RIGHT	LEFT	COMBINED	ESTIMATED SI
.5	100.91	85.90	93.40	2.50
1.0	39.85	29.83	34.84	2.35
2.0	15.15	10.65	12.90	1.96
4.1	6.92	4.47	5.69	1.59
8.1	2.93	1.90	2.41	1.67
16.2	1.15	.81	.98	1.91
32.4	.48	.38	.43	2.05
64.9	.22	.17	.20	2.05
129.8	.07	.07	.07	2.04

MRM ROUGHNESS (COUNTS/.2 MILE): 167.99

FLEXIBLE PAVEMENT SERVICEABILITY: 1.65

APPENDIX C

MRMCAL2--USER'S GUIDE
AND SOURCE LISTING

PROGRAM MRMCAL3 (INPUT,OUTPUT,TAPES=INPUT,TAPE6=OUTPUT)	MP	10
C	MR	20
C	MR	30
C M R M C A L 3	MR	40
C UPDATED: 8 MAY 1981	MR	50
C	MR	60
C	MR	70
C*****	MR	80
C	MR	90
C MRMCAL3 IS A PROGRAM USED TO CALIBRATE A MAYS ROAD METER AGAINST	MR	100
C AN AVAILABLE ROUGHNESS STATISTIC (MO) WITH WHICH IT ASSUMED TO	MR	110
C HAVE A LINEAR RELATIONSHIP:	MR	120
C	MR	130
C MPT = ALPHA + BETA * MO	MR	140
C	MR	150
C WHERE MPT IS THE PREDICTED RAW READING (COUNTS PER .2 MILE) OF	MR	160
C THE MAYS METER TO BE CALIBRATED AND MO IS THE PROFILE-DERIVED	MR	170
C MAYS METER SIMULATION USED AS THE CALIBRATION STANDARD.	MR	180
C	MR	190
C ONCE THE PARAMETERS ALPHA AND BETA ARE OBTAINED, THE	MR	200
C MAYS METER RAW READINGS, MT (COUNTS PER .2 MILE), CAN BE TRANS-	MR	210
C FORMED TO CALIBRATED VALUES BY	MR	220
C	MR	230
C MCI = (MPT - ALPHA) / BETA	MR	240
C	MR	250
C WHERE MCI IS A MAYS METER-DERIVED ESTIMATE OF MO (THE IDEAL	MR	260
C MAYS METER).	MR	270
C	MR	280
C FINALLY, THE CALIBRATED VALUES (MCI) CAN BE TRANSFORMED TO	MR	290
C SERVICEABILITY INDICES BY THE RELATION	MR	300
C	MR	310
C SIV = 5.0 * EXP(-(ALOG(32.0*MCI)/8.49331) ** 9.35661)	MR	320
C	MR	330
C	MR	340
C FOR MORE INFORMATION SEE CTR REPORT NO. 251-1.	MR	350
C	MR	360
C*****	MR	370
C	MR	380
C MRMCAL3 WAS WRITTEN IN APRIL 1981 BY PRENTISS RIDDLE AND DAVID	MR	390
C MCKENZIE OF THE CENTER FOR TRANSPORTATION RESEARCH AT THE	MR	400
C UNIVERSITY OF TEXAS AT AUSTIN.	MR	410
C	MR	420
C*****	MR	430
C*****	MR	440
C	MR	450
C INPUT GUIDE ==	MR	460
C	MR	470
C	MR	480
C AN INPUT DATA FILE CONSISTS OF ONE IDENTIFICATION CARD FOLLOWED BY	MR	490
C DATA FOR UP TO 35 SECTIONS.	MR	500
C	MR	510
C THE DATA FOR A SECTION CONSIST OF ONE SECTION HEADER CARD FOLLOWED	MR	520
C UP TO TEN RUN CARDS.	MR	530
C	MR	540
C THE FORMATS FOR THE VARIOUS SORTS OF CARD-IMAGES ARE AS FOLLOWS:	MR	550

C				MR	560
C				MR	570
C				MR	580
C				MR	590
C				MR	600
C				MR	610
C				MR	620
C				MR	630
C				MR	640
C				MR	650
C				MR	660
C				MR	670
C				MR	680
C				MR	690
C				MR	700
C				MR	710
C				MR	720
C				MR	730
C				MR	740
C				MR	750
C				MR	760
C				MR	770
C				MR	780
C				MR	790
C				MR	800
C				MR	810
C				MR	820
C				MR	830
C				MR	840
C				MR	850
C				MR	860
C				MR	870
C				MR	880
C				MR	890
C				MR	900
C				MR	910
C				MR	920
C				MR	930
C				MR	940
C				MR	950
C				MR	960
C				MR	970
C				MR	980
C				MR	990
C				MR	1000
C				MR	1010
C				MR	1020
C				MR	1030
C				MR	1040
C				MR	1050
C				MR	1060
C				MR	1070
C				MR	1080
C				MR	1090
C				MR	1100
C				MR	1110
C				MR	1120
C				MR	1130
C				MR	1140
C				MR	1150
C				MR	1160
C				MR	1170


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+           21HRESENT SERVICEABILITY/
+ 1H,41X,4MHINDEX: THE RELATIONSHIP BETWEEN MCI AND SIV WAS/
+ 1H,41X,4MHDERIVED INDEPENDENTLY OF THIS REGRESSION AND IS:/
+ 1H,41X,4MHSIV = 5.0*EXP(=(ALOG(32.0*MCI)/8.49331)**9.3566) .)
C
C ** CLEAR SUMMATION VARIABLES
SX = 0.0
SY = 0.0
XX = 0.0
YY = 0.0
XY = 0.0
XMAX = -BIG
YMAX = -BIG
XMIN = BIG
YMIN = BIG
C
C ** SUMMATION LOOP
DO 100 I=1,NSEC
  X0 = AMO(I)
  Y0 = AMY(I)
  XMIN = AMIN1 ( XMIN, X0 )
  YMIN = AMIN1 ( YMIN, Y0 )
  XMAX = AMAX1 ( XMAX, X0 )
  YMAX = AMAX1 ( YMAX, Y0 )
  SX = SX + X0
  SY = SY + Y0
  XX = XX + X0*X0
  YY = YY + Y0*Y0
  XY = XY + X0*Y0
100 CONTINUE
C
C ** CALCULATE REGRESSION COEFFICIENTS
DN = 1.0 / FLOAT( NSEC )
TOP = XY - SX*SY*DN
BOTTOM = XX - SX*SX*DN
RETA = TOP / BOTTOM
ALPHA = DN * ( SY - RETA*SX )
BETAP = 1.0 / RETA
ALPHAP = -ALPHA / RETA
SSR = RETA * TOP
SSE = YY - SY*SY*DN - SSR
SE = SQRT ( SSE / (NSEC-2) )
XMEAN = SX * DN
YMEAN = SY * DN
SESI = SE / SQRT(BOTTOM)
SEIN = SE * SQRT ( XX * DN / BOTTOM )
IF (NSEC .LT. 20) .OR. (NSEC .GT. 40) GO TO 110
CIA = T(NSEC-19) * SEIN
CTR = T(NSEC-19) * SESL
110 CONTINUE
RSQ = SSR / ( SSE + SSR )
C
C ** PRINT THE RESULTS
WRITE (6,200) NSEC,XMEAN,YMEAN
WRITE (6,210) ALPHA,RETA
IF (NSEC .GT. 20) .AND. (NSEC .LT. 40)
+   WRITE (6,220) ALPHA+CIA,ALPHA-CIA,RETA-CTR,RETA+CTR
WRITE (6,230) SEIN,SESL,RSQ,SE
WRITE (6,240) ALPHA,RETA,ALPHAP,BETAP
200 FORMAT (1H0,70X,69(1H*))//
+ 1H,54X,10HREGRESSION RESULTS//
+ 1H,44X,10HNUMBER OF SECTIONS:,9X,111/
LI 620
LI 630
LI 640
LI 650
LI 660
LI 670
LI 680
LI 690
LI 700
LI 710
LI 720
LI 730
LI 740
LI 750
LI 760
LI 770
LI 780
LI 790
LI 800
LI 810
LI 820
LI 830
LI 840
LI 850
LI 860
LI 870
LI 880
LI 890
LI 900
LI 910
LI 920
LI 930
LI 940
LI 950
LI 960
LI 970
LI 980
LI 990
LI 1000
LI 1010
LI 1020
LI 1030
LI 1040
LI 1050
LI 1060
LI 1070
LI 1080
LI 1090
LI 1100
LI 1110
LI 1120
LI 1130
LI 1140
LI 1150
LI 1160
LI 1170
LI 1180
LI 1190
LI 1200
LI 1210
LI 1220
LI 1230

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+ 1H .44X,8HMEAN MOI,20X,F11.2/ LI 1240
+ 1H .44X,8HMEAN MYI,20X,F11.2/ LI 1250
210 FORMAT (1H0,44X,6HAI PHA,22X,F11.4/ LI 1260
+ 1H .44X,5HRETA,23X,F11.4) LI 1270
220 FORMAT (1H0,44X,32H95 PERCENT CONFIDENCE INTERVALS:/ LI 1280
+ 1H .49X,FR,2,13H < ALPHA < ,FR,2/ LI 1290
+ 1H .49X,FR,2,13H < BETA < ,FR,2) LI 1300
230 FORMAT (1H0,44X,26HSTANDARD ERROR OF INTERCEPT, F11.4/ LI 1310
+ 1H .44X,26HSTANDARD ERROR OF SLOPE, 4X, F11.4/ LI 1320
+ 1H0,44X,10HRSQUARFD,18X,F11.5/ LI 1330
+ 1H .44X,26HSTANDARD ERROR OF RESIDUALS, F11.4) LI 1340
240 FORMAT (/1H0,44X,21HREGRESSION EQUATIONS:/ LI 1350
+ 1H0,47X,6HMP1 =,F11.4,3H +,F11.4,5H * MO/ LI 1360
+ 1H0,47X,6HMC1 =,F11.4,3H +,F11.4,5H * MY) LI 1370
C LI 1380
C LI 1390
RETURN LI 1400
END LI 1410

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SUBROUTINE PAGE ( NLINES ) PA 10
C PA 20
C ROUTINE TO SKIP TO THE NEXT PAGE AND PRINT A HEADING IF DESIRED. PA 30
C PA 40
C THE PARAMETER =NLINES= IS USED TO CONTROL PAGINATION. PA 50
C IF NLINES = 0, THE SUBROUTINE SKIPS TO A NEW PAGE AND PRINTS PA 60
C THE HEADER. OTHERWISE THE LINE-COUNTER =KLINES= IS INCREMENTED PA 70
C BY =NLINES= AND THE SUBROUTINE SKIPS TO A NEW PAGE ONLY IF PA 80
C THE RESULT IS GREATER THAN =MAXLNS=. PA 90
C PA 100
C***** PA 110
C PA 120
C9 IMPLICIT REAL*8 (A-H,O-Z) PA 130
COMMON /HEAD/ DATEM(4),DISFL,MRMID(4),IVELOC,DATEP(2),NPAGE,KLINES PA 140
DATA MAXLNS / 60 / PA 150
C PA 160
C PA 170
KLINES = KLINES + NLINES PA 180
IF ((NLINES .NE. 0) .AND. (KLINES .LE. MAXLNS)) RETURN PA 190
C PA 200
C ** PAGE EJECT PA 210
KLINES = NLINES + 9 PA 220
NPAGE = NPAGE + 1 PA 230
WRITE (6,100) NPAGE,DISFL,IVELOC,MRMID,DATEM,DATEP PA 240
100 FORMAT ( 1H1,13X,8HTRIM==>,72X,4NPAGE,I2,7X,8H<===TRIM/ PA 250
+ 1H .37X,20HTEXAS HIGHWAY DEPARTMENT, PA 260
+ 30H = MAYS RIDE METER CALIBRATION/ PA 270
+ 1H0,30X,60(1H*)/ PA 280
+ 1H .30X,15HDISTRICT/FILE: ,44,16X,21HCALIBRATION VELOCITY:, I3, PA 290
+ 4H MPH/ PA 300
+ 1H .30X,9HMRM NO.: ,444,10X,19HMAYS METER RUN: ,444/ PA 310
+ 1H .65X,19HDNOFILMETER RUN: ,244/ PA 320
+ 1H .30X,60(1H*)/ PA 330
C PA 340
RETURN PA 350
END PA 360

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SUBROUTINE P PLOT ( NSEC, AMO, AMI, ALPHA, BETA )          PP  10
C                                                         PP  20
C ROUTINE TO PRODUCE A PRINTER-PLOT OF MO VS MI AND THEIR PP  30
C REGRESSION LINE.                                       PP  40
C                                                         PP  50
C*****PP  60
C THE APPROACH USED TO CREATE THE PRINTER-PLOT IS AS FOLLOWS: PP  70
C                                                         PP  80
C THE ARRAY "SCREEN", 61 BY 37, IS A 6-RY-6-INCH CHARACTER BUFFER PP  90
C ON WHICH THE PLOT IS DRAWN. ITS X AND Y AXES CORRESPOND TO MO PP 100
C AND MI RESPECTIVELY. SCALE FACTORS ARE CHOSEN SO THAT THE RANGE PP 110
C OF MO VALUES OBSERVED MAPS ON TO THE 61 CELL-WIDTHS OF THE X PP 120
C AXIS AND THE RANGE OF MI VALUES OBSERVED MAPS ON TO THE 37 CELL- PP 130
C HEIGHTS OF THE Y-AXIS. PP 140
C                                                         PP 150
C FIRST, THE PROGRAM SELECTS THE ENDPONTS OF A SEGMENT OF THE PP 160
C REGRESSION LINE THAT CROSSES THE RANGE OF OBSERVED POINTS. THESE PP 170
C ENDPONTS ARE SCALED TO SCREEN COORDINATES AND THE LINE IS THEN PP 180
C DRAWN USING A "SIMPLE DIGITAL DIFFERENTIAL ANALYZER" ALGORITHM, PP 190
C OR DDA, AS DESCRIBED IN "PRINCIPLES OF INTERACTIVE GRAPHICS" BY PP 200
C NEWMAN AND SPROULL (PP. 24-25). PP 210
C                                                         PP 220
C NEXT, THE PROGRAM PLOTS THE OBSERVED (MO,MI) POINTS BY SIMPLY PP 230
C SCALING THEM TO SCREEN COORDINATES AND PLOTTING EACH POINT. PP 240
C                                                         PP 250
C FINALLY, "SCREEN" IS PRINTED OUT LINE-BY-LINE, TOP LINE FIRST, PP 260
C WITH THE AXES LABELLED WITH MO AND MI VALUES AT ONE-INCH PP 270
C INTERVALS. PP 280
C*****PP 290
C VARIABLES USED -- PP 300
C AMIDEL DIFFERENCE BETWEEN MAXIMUM AND MINIMUM VALUES OF MI PP 310
C AMIMAX MAXIMUM MI VALUE PP 320
C AMIMIN MINIMUM MI VALUE PP 330
C AMITIC DISTANCE BETWEEN TIC MARKS ON MI AXTS PP 340
C AMITCY MI TO Y CONVERSION FACTOR PP 350
C AMI1 MI VALUE FOR FIRST ENDPONIT OF REGRESSION LINE PP 360
C AMI2 MI VALUE FOR SECOND ENDPONIT OF REGRESSION LINE PP 370
C AMODEL DIFFERENCE BETWEEN MAXIMUM AND MINIMUM VALUES OF MO PP 380
C AMOMAX MAXIMUM MO VALUE PP 390
C AMOMIN MINIMUM MO VALUE PP 400
C AMOTIC DISTANCE BETWEEN TIC MARKS ON MO AXIS PP 410
C AMOTOX MO TO X CONVERSION FACTOR PP 420
C AMO1 MO VALUE FOR FIRST ENDPONIT OF REGRESSION LINE PP 430
C AMO2 MO VALUE FOR SECOND ENDPONIT OF REGRESSION LINE PP 440
C DELTA LENGTH ESTIMATE FOR REGRESSION LINE PP 450
C DELTAX X-AXIS LENGTH ESTIMATE FOR REGRESSION LINE PP 460
C DELTAY Y-AXIS LENGTH ESTIMATE FOR REGRESSION LINE PP 470
C INCH TIC-MARK COUNTER PP 480
C KBLANK CHARACTER BLANK (1H ) PP 490
C KO CHARACTER O (1H.) PP 500
C KSTAR CHARACTER STAR (1H*) PP 510
C NSTEPS NUMBER OF STEPS USED TO DRAW REGRESSION LINE PP 520
C SCREEN 61 BY 37 CHARACTER ARRAY USED FOR PLOT PP 530
C X X VALUE IN DDA PP 540
C XINC X INCREMENT IN DDA PP 550

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C      X1      X VALUE FOR FIRST ENDPOINT OF REGRESSION LINE      PP  590
C      X2      X VALUE FOR SECOND ENDPOINT OF REGRESSION LINE     PP  600
C      Y        Y VALUE IN DDA                                    PP  610
C      YINC     Y INCREMENT IN DDA                                PP  620
C      Y1      Y VALUE FOR FIRST ENDPOINT OF REGRESSION LINE     PP  630
C      Y2      Y VALUE FOR SECOND ENDPOINT OF REGRESSION LINE     PP  640
C
C*****PP  660
C
C      IMPLICIT REAL*8 (A-H,O-Z)                                  PP  680
C      DIMENSION AMO(NSEC), AMI(NSEC)                             PP  690
C      INTEGER SCREEN (61,37)                                     PP  700
C      DATA KBLANK,KO,KSTAR / 1H ,1H0,1H* /                      PP  710
C
C      CALL PAGE ( # )                                           PP  720
C
C      ** CLEAR SCREEN                                           PP  730
C      DO 110 I=1,61                                              PP  740
C          DO 100 J=1,37                                          PP  750
C              SCREEN (I,J) = KBLANK                             PP  760
C          CONTINUE                                             PP  770
100      CONTINUE                                               PP  780
110      CONTINUE                                               PP  790
C
C      ** DETERMINE SCALE FACTORS                                  PP  800
C      AMO(1) = AMO(1)                                           PP  810
C      AMOMAX = AMO(1)                                           PP  820
C      AMT(1) = AMI(1)                                           PP  830
C      AMIMIN = AMI(1)                                           PP  840
C      AMT(1) = AMI(1)                                           PP  850
C      AMIMAX = AMI(1)                                           PP  860
C      DO 200 K=2,NSEC                                           PP  870
C          IF (AMO(K) .LT. AMO(1)) AMO(1) = AMO(K)               PP  880
C          IF (AMO(K) .GT. AMOMAX) AMOMAX = AMO(K)               PP  890
C          IF (AMI(K) .LT. AMI(1)) AMI(1) = AMI(K)               PP  900
C          IF (AMI(K) .GT. AMIMAX) AMIMAX = AMI(K)               PP  910
200      CONTINUE                                               PP  920
C      AMODEL = AMOMAX - AMO(1)                                    PP  930
C      AMIDEL = AMT(1) - AMI(1)                                    PP  940
C
C
C      ** SELECT END POINT OF THE REGRESSION LINE TO BE DRAWN   PP  950
C      AMO1 = AMO(1)                                             PP  960
C      AMO2 = AMOMAX                                             PP  970
C      AMI1 = ALPHA + BETA*AMO1                                   PP  980
C      AMI2 = ALPHA + BETA*AMO2                                   PP  990
C
C      ** CONVERT TO SCREEN COORDINATES                           PP 1000
C      AMOTOX = 60.0 / AMODEL                                     PP 1010
C      AMITOX = 36.0 / AMIDEL                                     PP 1020
C      X1 = (AMO1 - AMO(1)) * AMOTOX                             PP 1030
C      Y2 = (AMO2 - AMO(1)) * AMOTOX                             PP 1040
C      Y1 = (AMI1 - AMI(1)) * AMITOX                             PP 1050
C      Y2 = (AMI2 - AMI(1)) * AMITOX                             PP 1060
C
C
C      ** SELECT X AND Y INCREMENTS FOR DDA                       PP 1070
C      DELTAX = X2 - X1                                          PP 1080
C      DELTAY = Y2 - Y1                                          PP 1090
C      DELTA = AMAXI ( ABS(DELTAX), ABS(DELTAY) )                 PP 1100
C      XINC = DELTAX / DELTA                                       PP 1110
C      YINC = DELTAY / DELTA                                       PP 1120
C
C
C      ** PERFORM THE ITERATIVE PART OF THE DDA PROCESS: STEP ALONG LINE PP 1130
C      X = Y1 + 1.5                                             PP 1140
C      Y = Y1 + 1.5                                             PP 1150
C      NSTEPS = INT(DELTA) + 5                                     PP 1160
C      DO 300 K=1,NSTEPS                                         PP 1170

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      I = INT(X1)
      J = INT(Y1)
C      ** PLOT POINT ON LINE (UNLESS CLIPPED)
      IF ((I.GE.1).AND.(I.LE.61).AND.(J.GE.1).AND.(J.LE.37))
+      SCREEN (I,J) = KSTAR
      X = X + XTNC
      Y = Y + YTNC
300 CONTINUE
C
C      ** PLOT (MO,MI) POINTS
      DO 400 K=1,NSEC
      I = INT( (AM0(K) = AM0MIN) * AMOT0X + 1.5 )
      J = INT( (AM1(K) = AM1MIN) * AM10Y + 1.5 )
      SCREEN (I,J) = K0
400 CONTINUE
C
C      ** PRINT THE SCREEN
      AMOTTC = 10.0 / 60.0 * AMO0EL
      AM1TTC = 6.0 / 36.0 * AM1DEL
      WRITE (6,520) ALPHA,BETA
      WRITE (6,530) AMTMAX,(SCREEN(I,37),I=1,61)
      DO 510 INCH=1,6
      DO 500 JJ=1,5
      I = 03 = 6*INCH = JJ
      WRITE (6,540) (SCREEN(I,J),I=1,61)
500 CONTINUE
      J = 37 = 6*INCH
      WRITE (6,530) AMTMAX=INCH*AMTTC,(SCREEN(I,J),I=1,61)
510 CONTINUE
      WRITE (6,550) (AM0MIN+(INCH-1)*AMOTTC,INCH=1,7)
520 FORMAT (/1H0,02X,45HPLOT OF MO VS. MI (INCLUDING REGRESSION LINE)/
+ 1H0,42X,7HALPHA =,F10,4,12X,6HRETA =,F10,4/
+ 1H0,36X,2H+;.6(10H+++++),1H+)
530 FORMAT (1H,36X,F6.1,1H;.61A1,1H)
540 FORMAT (1H,36X,1H+.61A1,1H)
550 FORMAT (1H,36X,2H+.6(10H+++++),1H+/
+ 1H,29X,7E10.1/
+ 1H0,40X,17HVERTICAL AXYS; MI,13X,19HHORIZONTAL AXIS; MO/
+ 1H0,54X,16H0 = DATA POINT/
+ 1H,54X,21H* = REGRESSION LINE)
C
      RETURN
      END

```

```

      SUBROUTINE READIN ( NSEC, KSECNO, AM0, AM1 )
C
C      ROUTINE TO READ IN AND ECHO-PRINT THE REGRESSION DATA.
C
C*****
C
C      IMPLICIT REAL*8 (A-H,O-Z)
C8      COMMON /HEAD/ DATEM(4),DISFL,MRMID(4),IVELOC,DATEP(2),NPAGE,KLINES
      DIMENSION KSECNO(1),AM0(1),AM1(1),RAWDAT(10,4),RUNTOT(10)
      LOGICAL THROU,CHART
      DATA KT / 1HT /
C
C      ** READ IDENTIFICATION INFORMATION

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READ (9,100) MRMID,DISPL,IVELOC,DATEM,DATEP,NRUNS,KTHROW,KCHART RE 140
100 FORMAT (4A4,X,A4,X,I2,X,4A4,X,2A4,X,I2,X,A1,X,A1) RE 150
C ** SET OUTLIER AND MRM UNITS FLAGS (=THROW= AND =CHART=) RE 160
THROW = .FALSE RE 170
IF (KTHROW .EQ. KT) THROW = .TRUE RE 180
CHART = .FALSE RE 190
IF (KCHART .EQ. KT) CHART = .TRUE RE 200
C ** SET DEFAULT VALUE OF NRUNS RE 210
IF (NRUNS .GT. 10) STOP 1 RE 220
IF ((NRUNS .EQ. 0) .AND. (.NOT. THROW)) NRUNS = 4 RE 230
IF ((NRUNS .EQ. 0) .AND. THROW) NRUNS = 5 RE 240
C ** SET DEFAULT VALUE OF IVELOC RE 250
IF (IVELOC .EQ. 0) IVELOC = 50 RE 260
C RE 270
C ** PRINT PAGE HEADER RE 280
CALL PAGE ( 0 ) RE 290
CALL PAGE ( 0 ) RE 300
WRITE (6,150) NRUNS RE 310
IF (THROW) WRITE (6,160) RE 320
IF (.NOT. THROW) WRITE (6,170) RE 330
IF (CHART) WRITE (6,180) RE 340
IF (.NOT. CHART) WRITE (6,190) RE 350
150 FORMAT (/1H0,53X,21HI N P U T D A T A / RE 360
+ 1H0,38X,9HTHERE ARE,13,18H RUNS PER SECTION.) RE 370
160 FORMAT (1H ,38X,37HONE OUTLYING RUN PER SECTION WILL BE , RE 380
+ 10HDISCARDED.) RE 390
170 FORMAT (1H ,38X,36HOUTLYING RUNS WILL NOT BE DISCARDED.) RE 400
180 FORMAT (1H ,38X,40HMAYS METER INPUT UNITS ARE CHART INCHES./) RE 410
190 FORMAT (1H ,38X,34HMAYS METER INPUT UNITS ARE COUNTS./) RE 420
C RE 430
C ** LOOP TO READ AND PROCESS DATA SECTION BY SECTION RE 440
NSEC = 0 RE 450
200 CONTINUE RE 460
NSEC = NSEC + 1 RE 470
READ (5,210,END=400) KSECNO(NSEC),AMO(NSEC), RE 480
+ ((RAWDAT(I,J),J=1,4),I=1,NRUNS) RE 490
210 FORMAT (14,F10.0)/(4F10.0) RE 500
C RE 510
C ** CALCULATE THE TOTAL MAYS METER READING FOR EACH RUN RE 520
DO 220 I=1,NRUNS RE 530
RUNTOT(I) = RAWDAT(I,1)+RAWDAT(I,2)+RAWDAT(I,3)+RAWDAT(I,4) RE 540
220 CONTINUE RE 550
C RE 560
C ** CALCULATE AMI(NSEC), ELIMINATING THE OUTLYING RUN IF NECESSARY RE 570
TOT = 0.0 RE 580
DO 230 I=1,NRUNS RE 590
TOT = TOT + RUNTOT(I) RE 600
230 CONTINUE RE 610
TEMPMI = TOT / FLOAT(NRUNS) RE 620
NOUTL = 0 RE 630
IF (.NOT. THROW) GO TO 250 RE 640
NOUTL = 1 RE 650
DO 240 I=2,NRUNS RE 660
IF (ABS(RUNTOT(I)-TEMPMI) .GT. ABS(RUNTOT(NOUTL)-TEMPMI)) RE 670
+ NOUTL = I RE 680
240 CONTINUE RE 690
TEMPMI = (TOT - RUNTOT(NOUTL)) / 4.0 RE 700
250 AMI(NSEC) = TEMPMI RE 710
IF (CHART) AMI(NSEC) = AMI(NSEC) * 32.0 RE 720
C RE 730
C ** PRINT DATA FOR SECTION RE 740
CALL PAGE ( NRUNS+3 ) RE 750

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300      WRITE (6,100) KSECNO(NSEC),AMO(NSEC),AMI(NSEC)
      FORMAT (1H0,38X,14HSECTION NUMBER,14,8H MO =,F8.2,
+         AM MI =,F8.2/)
      DO 330 I=1,NMIINS
310      WRITE (6,310) (RANDAT(I,J),J=1,4),RUNTOT(I)
      FORMAT (51X,3(F6.1,2H +),F6.1,2H =,F6.1)
      IF (I.EQ. NOUTL) WRITE (6,320)
320      FORMAT (1H+,89X,9H(OUTLIER))
330      CONTINUE
C
C      ** END-OF-LOOP
      IF (NSEC.LT. 35) GO TO 200
C
C      ** TOO MANY SECTIONS
      WRITE (6,350) NSEC
350      FORMAT (1H0,38X,4HONLY,13,17H SECTIONS READ ==/
+         1H,38X,32MHY ADDITIONAL SECTIONS IGNORED.)
      RETURN
C
C      ** NORMAL LOOP EXIT (END-OF-FILE)
400      CONTINUE
      NSEC = NSEC - 1
      RETURN
      END

```

```

SUBROUTINE RSTAR ( NSEC, KSECNO, AMO, AMI, ALPHA, BETA )
C
C      ROUTINE TO PRINT THE RESIDUAL TABLE == SECTION NUMBER,
C      MI, MO, MCI, AND MCI-MO FOR EACH SECTION.
C
C*****
C
C      IMPLICIT REAL*8 (A-H,O-Z)
      DIMENSION KSECNO(1),AMO(1),AMI(1)
C
      CALL PAGE(0)
C
      WRITE (6,100)
100      FORMAT (/1H0,57X,14HRESIDUAL TABLE/
+ 1H0,37X,4HSECTION MAYS METER EXPECTED PREDIC,
+ 14H,38X,6HNUMBER,6X,4H(MI),9X,2HMO,6X,9HMO (MCI),6X,2HMO/)
C
      DO 300 I=1,NSEC
          AMCT = (AMI(I) - ALPHA) / BETA
          RES = AMCT - AMO(I)
          WRITE (6,200) KSECNO(I),AMI(I),AMO(I),AMCI,RES
200          FORMAT (1H,38X,14,F12.1,F12.1,F12.1,F11.1)
300      CONTINUE
C
      RETURN
      END

```

TEXAS HIGHWAY DEPARTMENT - MAYS RYDE METER CALIBRATION

 DISTRICT/FILES: 7 CALIBRATION VELOCITY: 50 MPH
 MRM NO.: 07-9873-A MAYS METER RUNS: 01/23-25/80
 PROFILOMETER RUNS: JAN 80

I N P U T D A T A :

THERE ARE 4 RUNS PER SECTION.
 OUTLYING RUNS WILL NOT BE DISCARDED.
 MAYS METER INPUT UNITS ARE COUNTS.

SECTION NUMBER	19	MO =	69.22	MI =	29.00
	3.0 +	6.0 +	13.0 +	8.0 =	30.0
	6.0 +	7.0 +	10.0 +	5.0 =	28.0
	4.0 +	6.0 +	15.0 +	3.0 =	28.0
	2.0 +	8.0 +	14.0 +	6.0 =	30.0
SECTION NUMBER	45	MO =	140.11	MI =	263.50
	73.0 +	71.0 +	77.0 +	26.0 =	247.0
	69.0 +	82.0 +	71.0 +	36.0 =	258.0
	73.0 +	87.0 +	84.0 +	33.0 =	277.0
	77.0 +	83.0 +	84.0 +	28.0 =	272.0
SECTION NUMBER	44	MO =	210.00	MI =	104.75
	31.0 +	29.0 +	25.0 +	23.0 =	108.0
	28.0 +	29.0 +	24.0 +	24.0 =	105.0
	27.0 +	25.0 +	27.0 +	22.0 =	101.0
	29.0 +	31.0 +	24.0 +	21.0 =	105.0
SECTION NUMBER	41	MO =	129.16	MI =	143.50
	39.0 +	36.0 +	40.0 +	28.0 =	143.0
	37.0 +	38.0 +	36.0 +	31.0 =	142.0
	37.0 +	37.0 +	39.0 +	33.0 =	146.0
	42.0 +	31.0 +	37.0 +	33.0 =	143.0
SECTION NUMBER	43	MO =	150.45	MI =	112.00
	20.0 +	25.0 +	34.0 +	30.0 =	109.0
	26.0 +	24.0 +	39.0 +	27.0 =	116.0
	23.0 +	22.0 +	37.0 +	31.0 =	113.0
	23.0 +	19.0 +	38.0 +	30.0 =	110.0
SECTION NUMBER	9	MO =	59.20	MI =	36.50
	4.0 +	7.0 +	14.0 +	10.0 =	35.0
	9.0 +	9.0 +	10.0 +	10.0 =	38.0
	11.0 +	7.0 +	13.0 +	8.0 =	39.0
	11.0 +	8.0 +	7.0 +	8.0 =	34.0

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TEXAS HIGHWAY DEPARTMENT - MAYS RIDE METER CALIBRATION

 DISTRICT/FILE: 7 CALIBRATION VELOCITY: 50 MPH
 MRM NO.: 07-9873-A MAYS METER RUN: 01/23-25/80
 PROFILEMETER RUN: JAN 80

SECTION NUMBER 40 MO = 55.56 MI = 25.00

6.0 + 8.0 + 5.0 + 7.0 = 26.0
 10.0 + 4.0 + 6.0 + 5.0 = 25.0
 7.0 + 3.0 + 4.0 + 7.0 = 21.0
 8.0 + 5.0 + 4.0 + 11.0 = 28.0

SECTION NUMBER 39 MO = 215.30 MI = 215.25

44.0 + 32.0 + 42.0 + 50.0 = 216.0
 41.0 + 31.0 + 78.0 + 60.0 = 210.0
 43.0 + 35.0 + 76.0 + 56.0 = 210.0
 51.0 + 33.0 + 79.0 + 62.0 = 225.0

SECTION NUMBER 38 MO = 123.59 MI = 98.75

27.0 + 21.0 + 27.0 + 26.0 = 101.0
 23.0 + 29.0 + 24.0 + 21.0 = 97.0
 25.0 + 27.0 + 27.0 + 20.0 = 99.0
 20.0 + 22.0 + 28.0 + 20.0 = 90.0

SECTION NUMBER 12 MO = 64.45 MI = 35.00

11.0 + 6.0 + 7.0 + 9.0 = 33.0
 11.0 + 7.0 + 5.0 + 15.0 = 38.0
 11.0 + 7.0 + 5.0 + 10.0 = 33.0
 12.0 + 9.0 + 6.0 + 9.0 = 36.0

SECTION NUMBER 32 MO = 22.44 MI = 11.00

3.0 + 5.0 + 1.0 + 2.0 = 11.0
 1.0 + 3.0 + 1.0 + 4.0 = 9.0
 1.0 + 5.0 + 2.0 + 4.0 = 12.0
 1.0 + 4.0 + 1.0 + 6.0 = 12.0

SECTION NUMBER 10 MO = 35.49 MI = 18.50

3.0 + 3.0 + 8.0 + 7.0 = 21.0
 2.0 + 8.0 + 2.0 + 5.0 = 17.0
 1.0 + 5.0 + 8.0 + 4.0 = 18.0
 1.0 + 3.0 + 6.0 + 8.0 = 18.0

SECTION NUMBER 13 MO = 103.28 MI = 84.50

17.0 + 16.0 + 17.0 + 33.0 = 83.0
 15.0 + 17.0 + 19.0 + 37.0 = 88.0
 15.0 + 17.0 + 19.0 + 31.0 = 82.0
 20.0 + 16.0 + 20.0 + 29.0 = 85.0

TEXAS HIGHWAY DEPARTMENT - MAYS RIDGE METER CALIBRATION

 DISTRICT/FILE: 7 CALIBRATION VELOCITY: 50 MPH
 MPM NO.: 07-9873-A MAYS METER RUN: 01/23-25/80
 PROFILEMETER RUN: JAN 80

SECTION NUMBER 14 MO = 65.66 MI = 37.00

12.0 + 10.0 + 9.0 + 6.0 = 37.0
 13.0 + 10.0 + 9.0 + 6.0 = 38.0
 13.0 + 9.0 + 6.0 + 7.0 = 35.0
 12.0 + 8.0 + 9.0 + 9.0 = 38.0

SECTION NUMBER 15 MO = 64.99 MI = 46.25

10.0 + 12.0 + 8.0 + 9.0 = 47.0
 16.0 + 10.0 + 6.0 + 7.0 = 47.0
 20.0 + 13.0 + 6.0 + 14.0 = 53.0
 21.0 + 11.0 + 8.0 + 6.0 = 46.0

SECTION NUMBER 37 MO = 110.85 MI = 85.50

28.0 + 21.0 + 14.0 + 12.0 = 75.0
 32.0 + 26.0 + 13.0 + 29.0 = 100.0
 26.0 + 28.0 + 13.0 + 21.0 = 88.0
 30.0 + 23.0 + 16.0 + 16.0 = 79.0

SECTION NUMBER 36 MO = 53.19 MI = 40.25

8.0 + 11.0 + 12.0 + 12.0 = 43.0
 8.0 + 16.0 + 7.0 + 8.0 = 39.0
 9.0 + 12.0 + 12.0 + 10.0 = 43.0
 10.0 + 8.0 + 9.0 + 9.0 = 36.0

SECTION NUMBER 35 MO = 195.26 MI = 140.50

31.0 + 35.0 + 37.0 + 30.0 = 133.0
 28.0 + 30.0 + 34.0 + 40.0 = 140.0
 27.0 + 39.0 + 36.0 + 39.0 = 141.0
 31.0 + 37.0 + 39.0 + 37.0 = 144.0

SECTION NUMBER A MO = 59.03 MI = 53.75

10.0 + 15.0 + 11.0 + 8.0 = 52.0
 16.0 + 14.0 + 14.0 + 10.0 = 54.0
 16.0 + 13.0 + 11.0 + 11.0 = 51.0
 17.0 + 14.0 + 12.0 + 15.0 = 58.0

SECTION NUMBER S MO = 18.53 MI = 14.25

2.0 + 3.0 + 3.0 + 7.0 = 15.0
 3.0 + 2.0 + 3.0 + 6.0 = 14.0
 6.0 + 3.0 + 2.0 + 3.0 = 14.0
 4.0 + 3.0 + 1.0 + 6.0 = 14.0

TEXAS HIGHWAY DEPARTMENT - MAYS RIDE METER CALIBRATION

PAGE 4

 DISTRICT/FILE: 7 CALIBRATION VELOCITY: 50 MPH
 MRM NO.: 07-9873-A MAYS METER RUN: 01/23-25/80
 PROFILEMETER RUN: JAN 80

SECTION NUMBER 3 MO = 52.87 MI = 32.75

14.0 + 6.0 + 6.0 + 7.0 = 33.0
 6.0 + 13.0 + 5.0 + 6.0 = 30.0
 7.0 + 10.0 + 5.0 + 9.0 = 31.0
 10.0 + 13.0 + 5.0 + 9.0 = 37.0

SECTION NUMBER 6 MO = 103.42 MI = 68.00

20.0 + 15.0 + 22.0 + 16.0 = 73.0
 21.0 + 14.0 + 17.0 + 12.0 = 64.0
 23.0 + 13.0 + 17.0 + 14.0 = 67.0
 23.0 + 13.0 + 16.0 + 16.0 = 68.0

SECTION NUMBER 2 MO = 185.56 MI = 130.25

35.0 + 27.0 + 27.0 + 42.0 = 131.0
 35.0 + 29.0 + 20.0 + 41.0 = 134.0
 36.0 + 27.0 + 21.0 + 40.0 = 124.0
 36.0 + 33.0 + 21.0 + 42.0 = 132.0

SECTION NUMBER 7 MO = 11.83 MI = 10.50

2.0 + 3.0 + 2.0 + 1.0 = 8.0
 2.0 + 2.0 + 1.0 + 5.0 = 10.0
 2.0 + 3.0 + 2.0 + 4.0 = 11.0
 4.0 + 2.0 + 3.0 + 4.0 = 13.0

SECTION NUMBER 23 MO = 102.89 MI = 73.75

31.0 + 12.0 + 14.0 + 16.0 = 73.0
 31.0 + 16.0 + 15.0 + 13.0 = 75.0
 20.0 + 14.0 + 15.0 + 16.0 = 75.0
 32.0 + 12.0 + 16.0 + 14.0 = 74.0

SECTION NUMBER 1 MO = 85.05 MI = 49.50

11.0 + 11.0 + 9.0 + 20.0 = 51.0
 10.0 + 13.0 + 8.0 + 14.0 = 45.0
 12.0 + 13.0 + 9.0 + 19.0 = 53.0
 12.0 + 11.0 + 8.0 + 18.0 = 49.0

SECTION NUMBER 28 MO = 45.74 MI = 30.25

8.0 + 5.0 + 7.0 + 12.0 = 32.0
 7.0 + 4.0 + 10.0 + 10.0 = 31.0
 5.0 + 3.0 + 7.0 + 13.0 = 28.0
 7.0 + 5.0 + 8.0 + 10.0 = 30.0

TEXAS HIGHWAY DEPARTMENT - MAYS RIDE METER CALIBRATION

```

*****
DISTRICT/FILE: 7 CALIBRATION VELOCITY: 50 MPH
MRM NO.: 07-9873-A MAYS METER RUN: 01/23-25/80
PROFILOMETER RUN: JAN 80
*****

```

REGRESSION VARIABLES:

```

MO PROFILE-DERIVED MAYS METER SIMULATION USED AS THE
CALIBRATION STANDARD (COUNTS PER .2 MILE).

MT ACTUAL RAW READINGS OF THE MAYS METER TO BE CALIBRATED
(COUNTS PER .2 MILE).

MPI PREDICTED MI BASED ON MO.

MCI PREDICTED MO BASED ON MT (CALIBRATED MAYS METER
READING).

SIV RESCALING OF MCI TO PREDICT THE PRESENT SERVICEABILITY
INDEX. THE RELATIONSHIP BETWEEN MCI AND SIV WAS
DERIVED INDEPENDENTLY OF THIS REGRESSION AND IS:

```

$$SIV = 5.0 * EXP(-(ALOG(12.0 * MCI) / 8.4933) ** 9.3566)$$

REGRESSION RESULTS:

```

NUMBER OF SECTIONS: 27
MEAN MO: 101.57
MEAN MI: 73.77

```

```

ALPHA: -6.6251
BETA: .7915

```

```

95 PERCENT CONFIDENCE INTERVALS:
-19.62 < ALPHA < 6.37
.69 < BETA < .89

```

```

STANDARD ERROR OF INTERCEPT: 6.3333
STANDARD ERROR OF SLOPE: .0505

```

```

R-SQUARED: .90778
STANDARD ERROR OF RESIDUALS: 19.3363

```

REGRESSION EQUATIONS:

$$MPI = -6.6251 + .7915 * MO$$

$$MCI = 1.3706 + 1.2635 * MI$$

TEXAS HIGHWAY DEPARTMENT - MAYS RIDE METER CALIBRATION PAGE 6

DISTRICT/FILE: 7 CALIBRATION VELOCITY: 50 MPH
MRM NO.: 07-9873-A MAYS METER RUN: 01/23-25/80
PROFILOMETER RUN: JAN 80

CALIBRATION TABLE

MAYS METER (MI)	PREDICTED MO (MCI)	SIV
259,9	336,8	.5
239,1	310,4	.6
221,4	288,3	.7
206,3	269,2	.8
193,2	252,2	.9
181,3	237,4	1,0
170,8	223,2	1,1
161,2	212,0	1,2
152,3	200,8	1,3
144,1	190,4	1,4
136,4	180,0	1,5
129,4	171,0	1,6
122,8	163,4	1,7
116,5	154,4	1,8
110,6	148,1	1,9
105,0	141,0	2,0
99,6	134,2	2,1
94,5	127,4	2,2
89,6	121,4	2,3
85,0	115,7	2,4
80,5	110,1	2,5
76,2	104,7	2,6
72,1	99,4	2,7
68,0	94,3	2,8
64,2	89,8	2,9
60,4	84,7	3,0
56,8	80,1	3,1
53,2	75,4	3,2
49,4	71,3	3,3
46,4	67,0	3,4
43,2	62,9	3,5
39,9	58,8	3,6
36,8	54,0	3,7
33,7	51,0	3,8
30,7	47,1	3,9
27,7	43,4	4,0
24,7	39,6	4,1
21,8	34,0	4,2
18,9	32,2	4,3
16,8	28,5	4,4
13,8	24,8	4,5
10,8	21,1	4,6
7,8	17,2	4,7
3,7	13,0	4,8
.8	8,4	4,9
-6.6	.8	5,0

TEXAS HIGHWAY DEPARTMENT - MAYS RIDE METER CALIBRATION

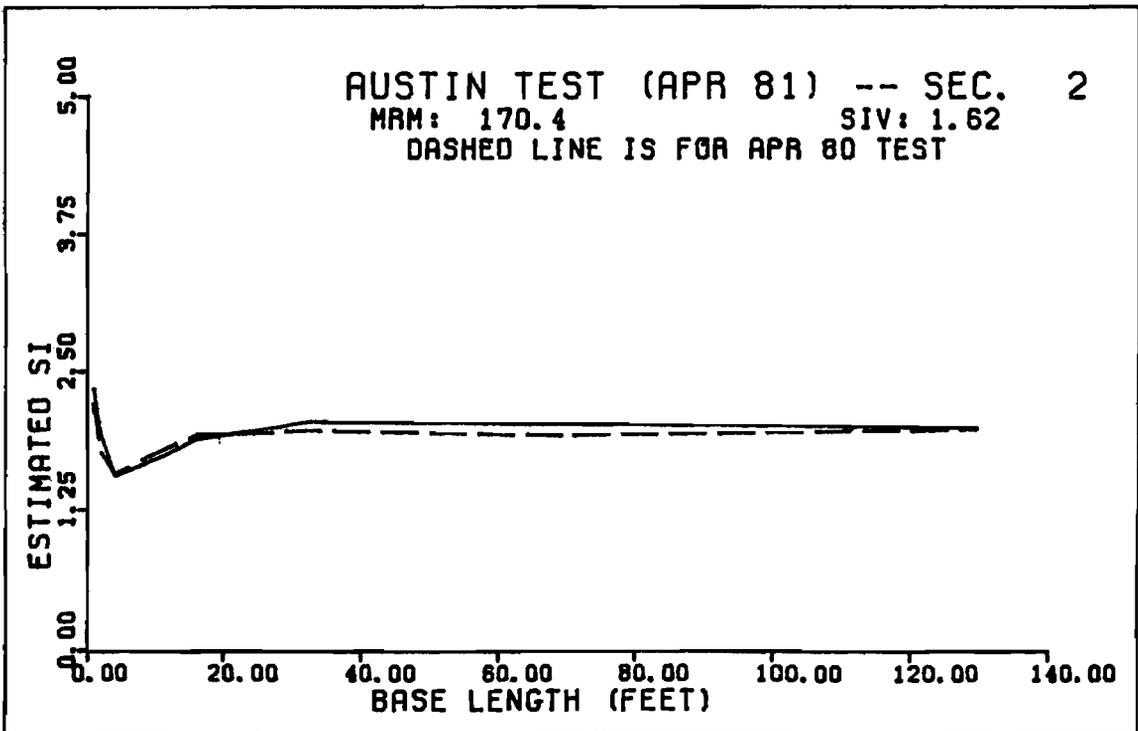
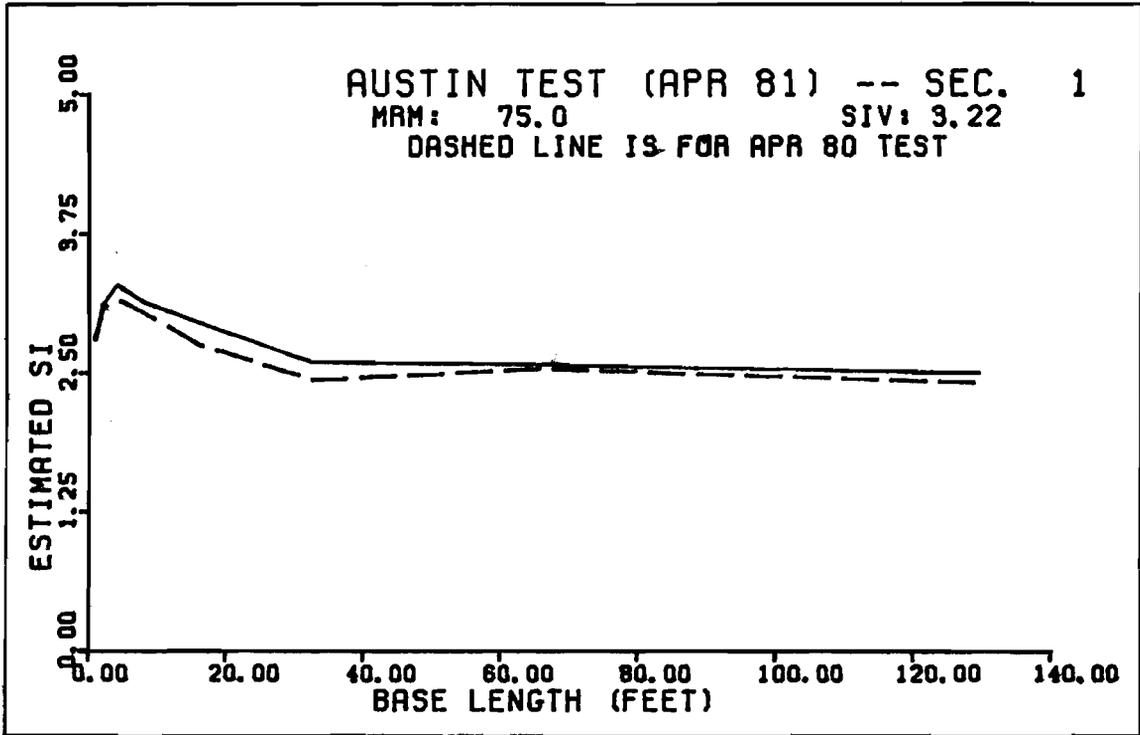
 DISTRICT/FILE: 7 CALIBRATION VELOCITY: 50 MPH
 MRM NO.: 07-9873-A MAYS METER RUN: 01/23-25/80
 PROFILEMETER RUN: JAN 80

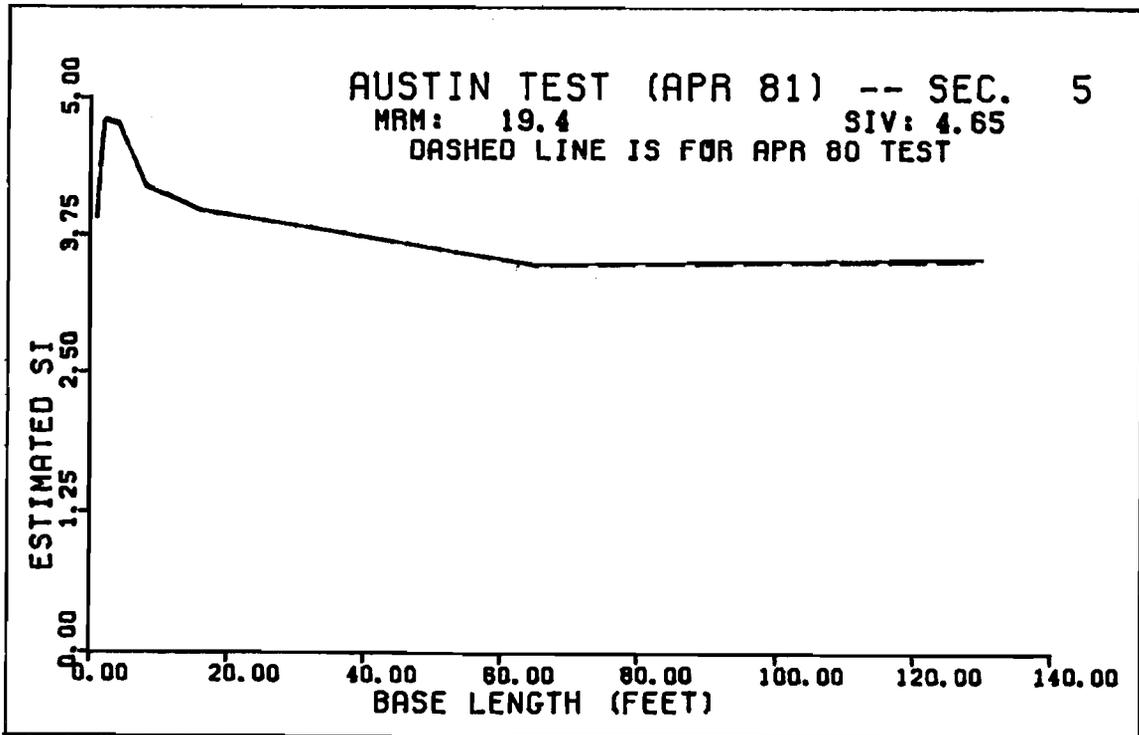
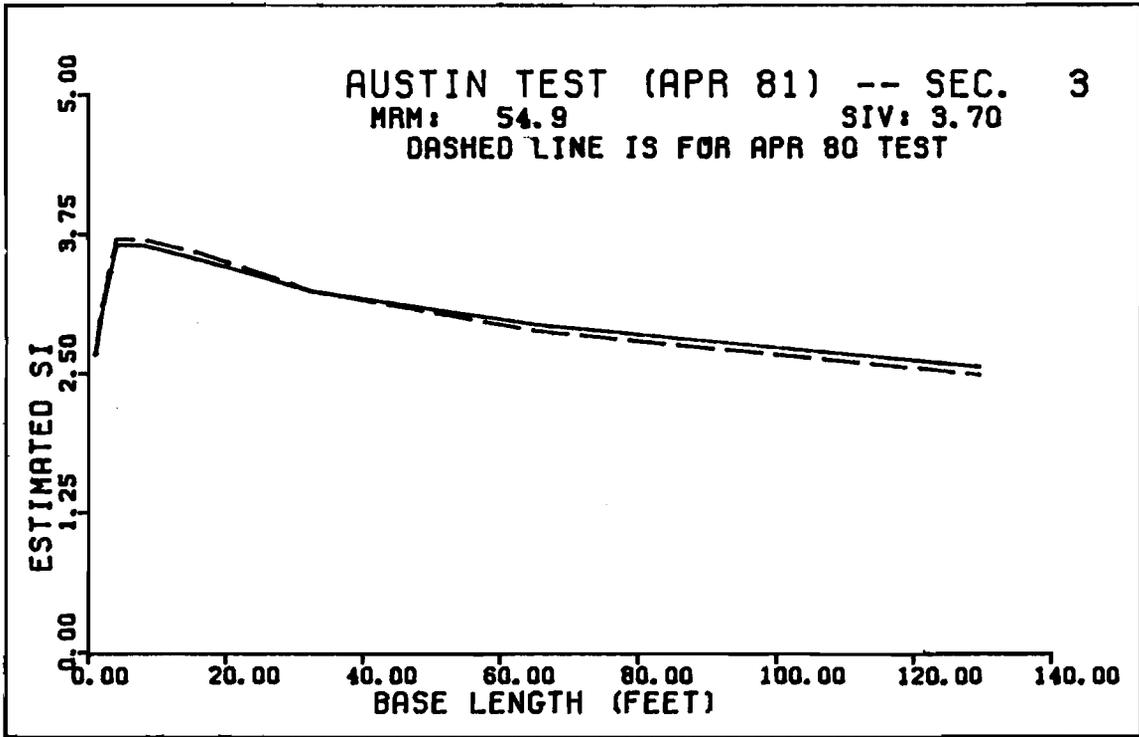
RESIDUAL TABLE

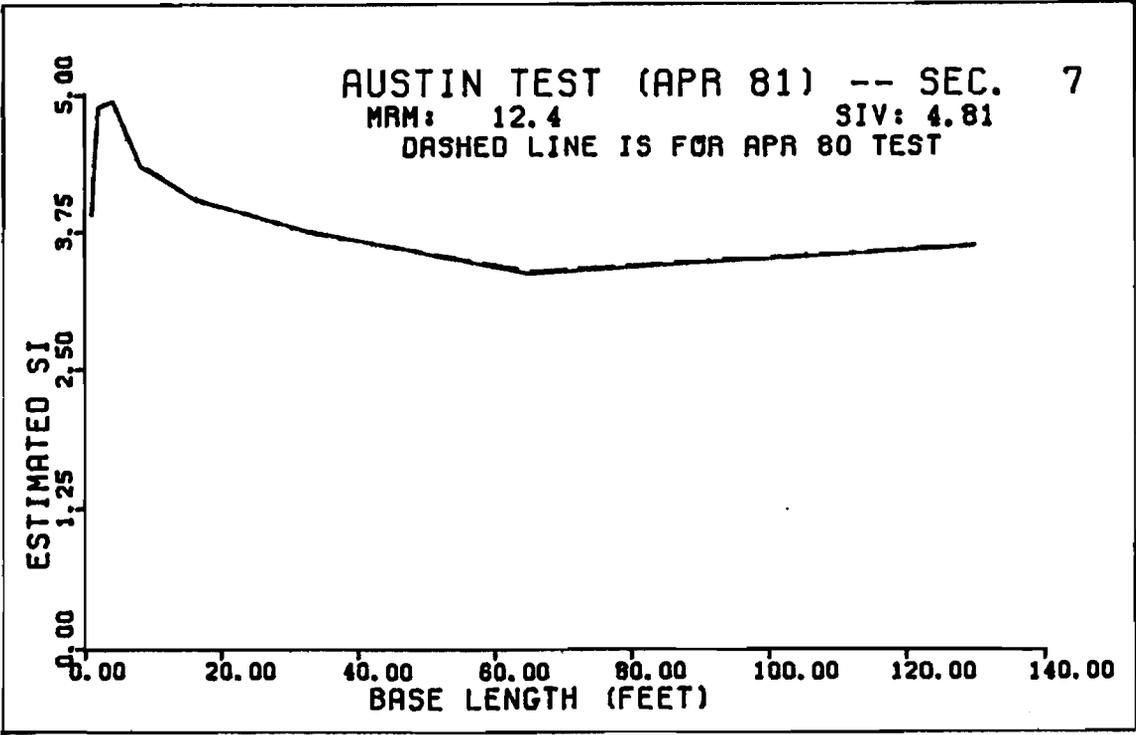
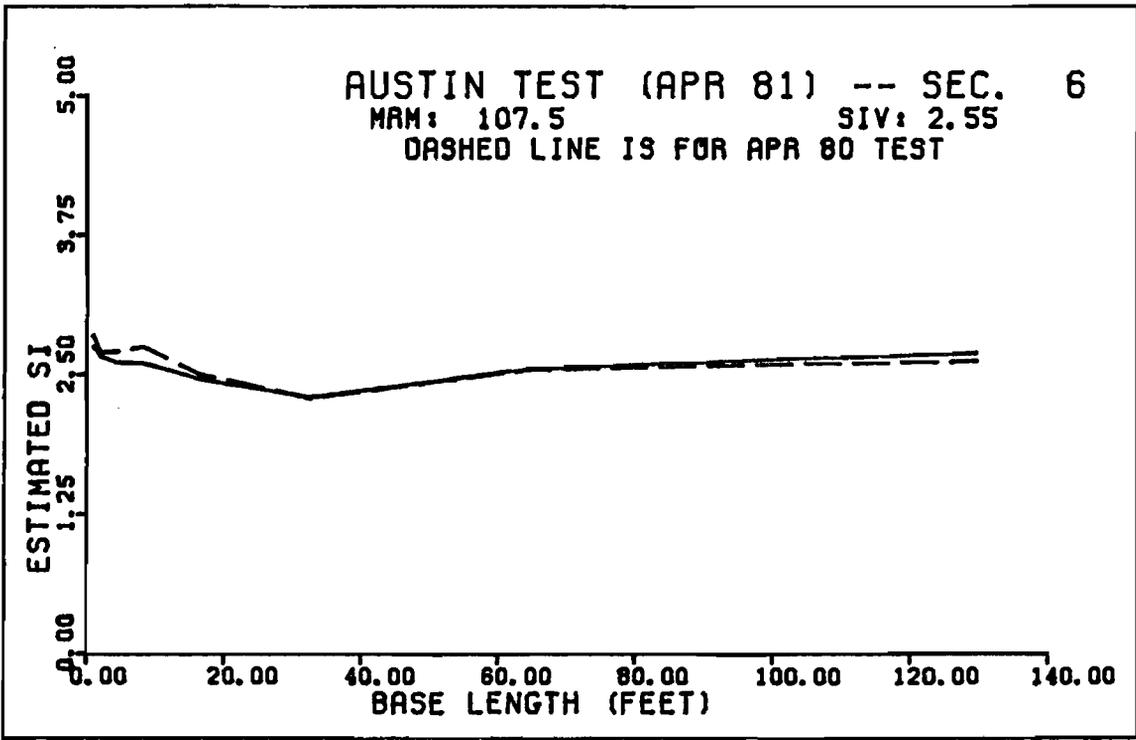
SECTION NUMBER	MAYS METER (MI)	EXPECTED MO	PREDICTED MO (MCI)	RESIDUAL MO
19	29.0	69.2	45.0	-24.2
45	263.5	300.1	301.3	1.2
44	100.0	210.9	140.7	-70.2
41	143.5	129.2	149.7	60.5
43	112.0	154.9	149.9	-5.0
9	36.5	59.2	54.5	-4.7
40	25.0	55.6	40.0	-15.6
39	215.3	218.3	280.3	62.0
38	98.0	123.6	133.1	9.5
12	35.0	44.4	52.6	-11.9
32	11.0	22.6	22.3	-.4
10	18.5	35.5	31.7	-3.7
13	84.5	103.3	115.1	11.9
14	37.0	65.7	55.1	-10.5
15	48.3	65.0	69.3	4.3
37	65.5	110.8	116.4	5.5
36	40.3	53.2	59.2	6.0
35	140.5	195.3	185.9	-9.4
8	53.0	59.0	76.3	17.3
5	14.3	18.5	26.4	7.8
3	32.0	42.9	49.7	-3.1
6	68.0	103.4	94.3	-9.1
2	130.3	185.6	172.9	-12.6
7	10.5	11.8	21.6	9.8
23	73.8	102.9	101.6	-1.3
1	49.5	85.9	70.9	-15.0
28	30.3	45.7	46.6	.9

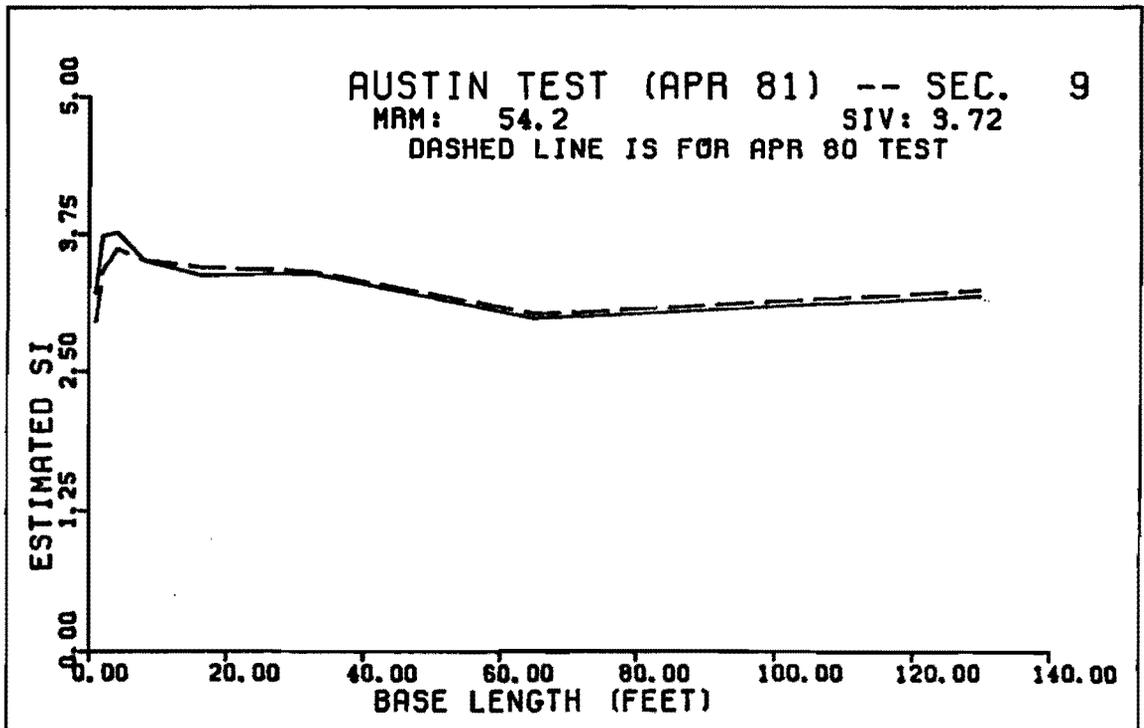
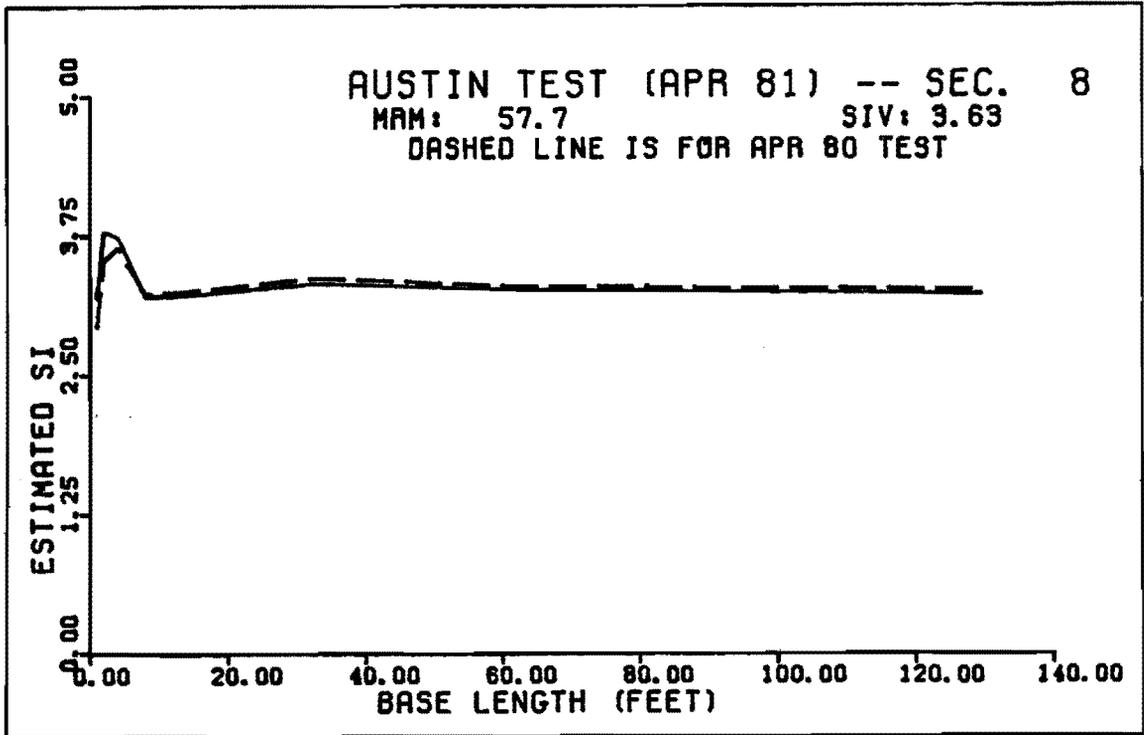
APPENDIX D

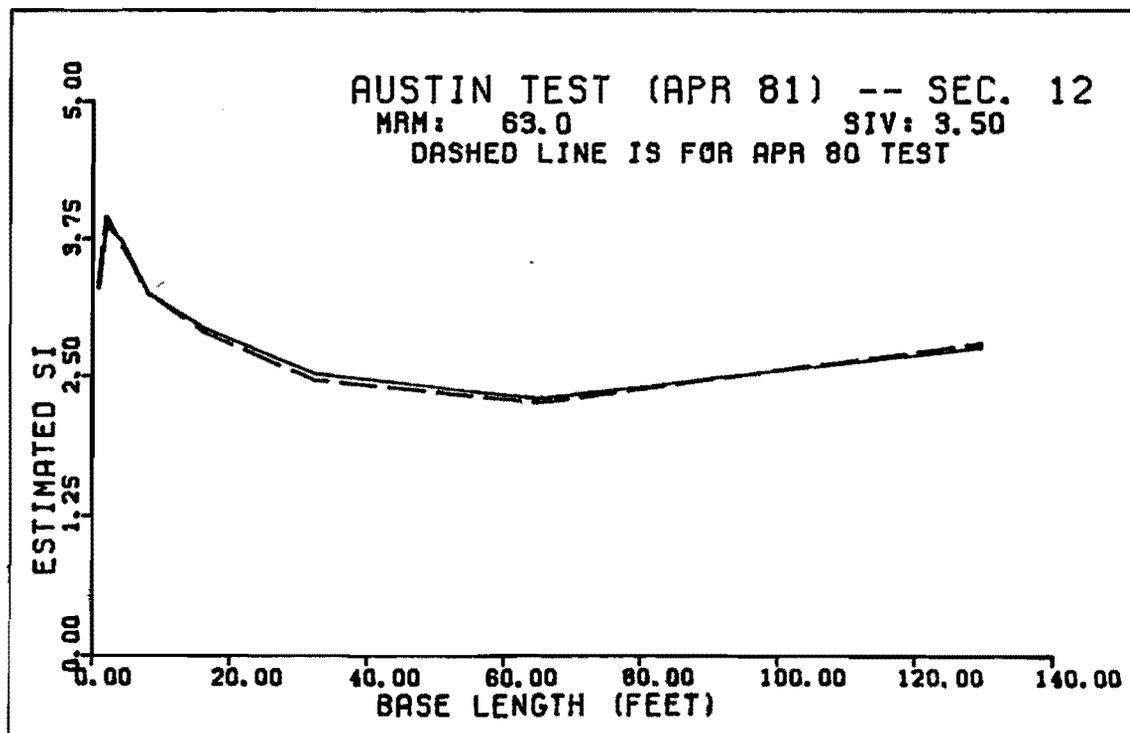
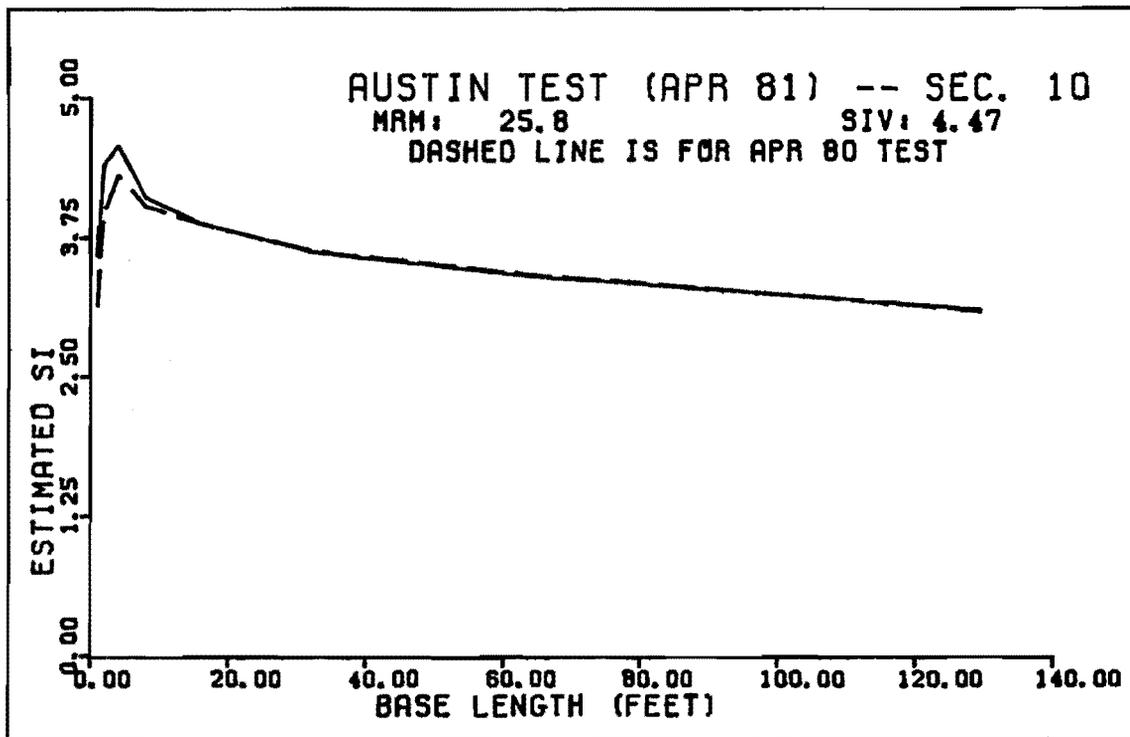
RMSVA ROUGHNESS SIGNATURES OF THE
AUSTIN TEST SECTIONS

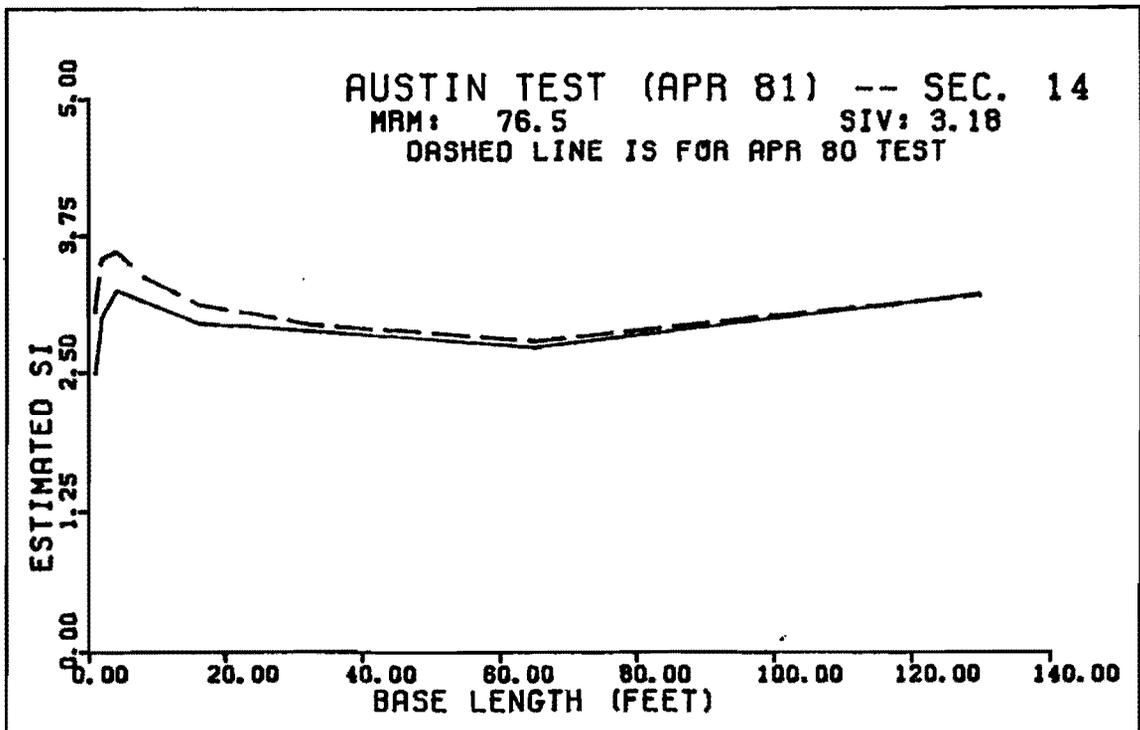
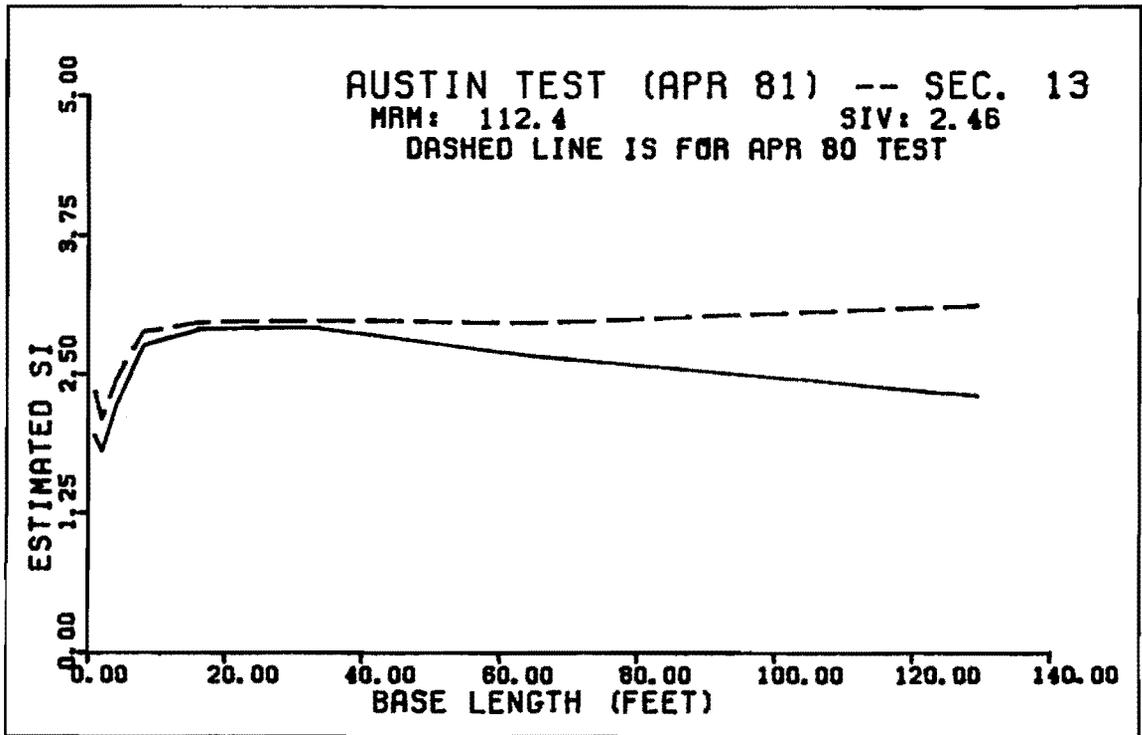


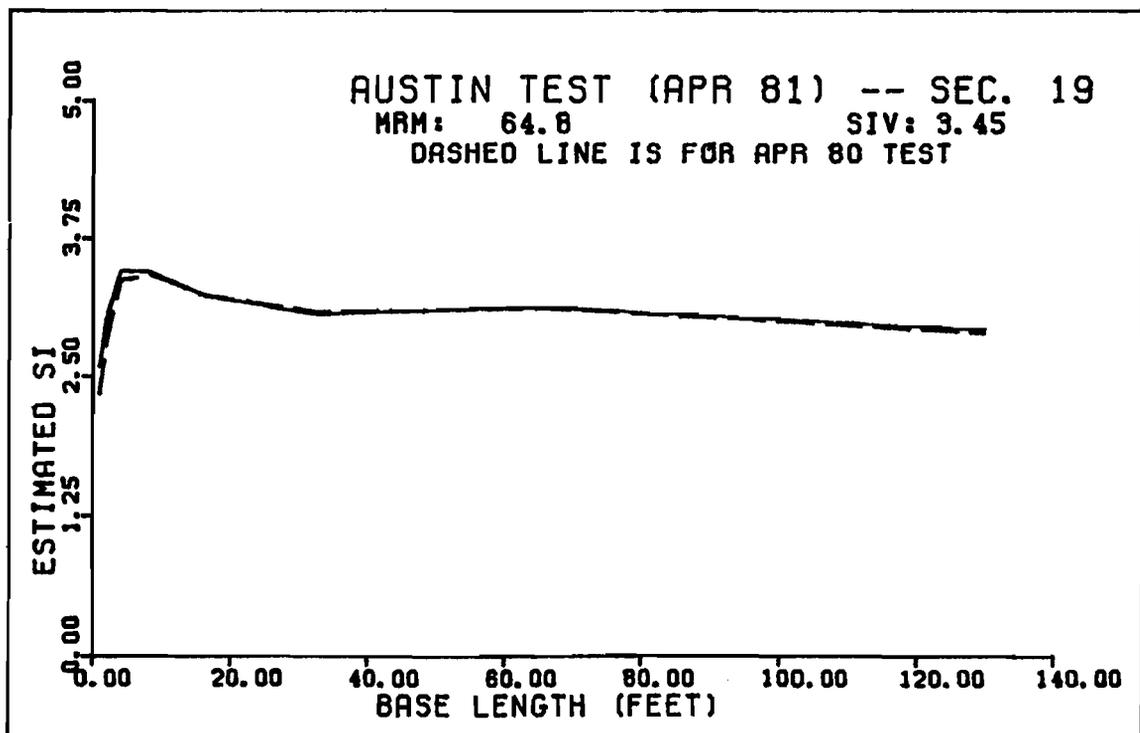
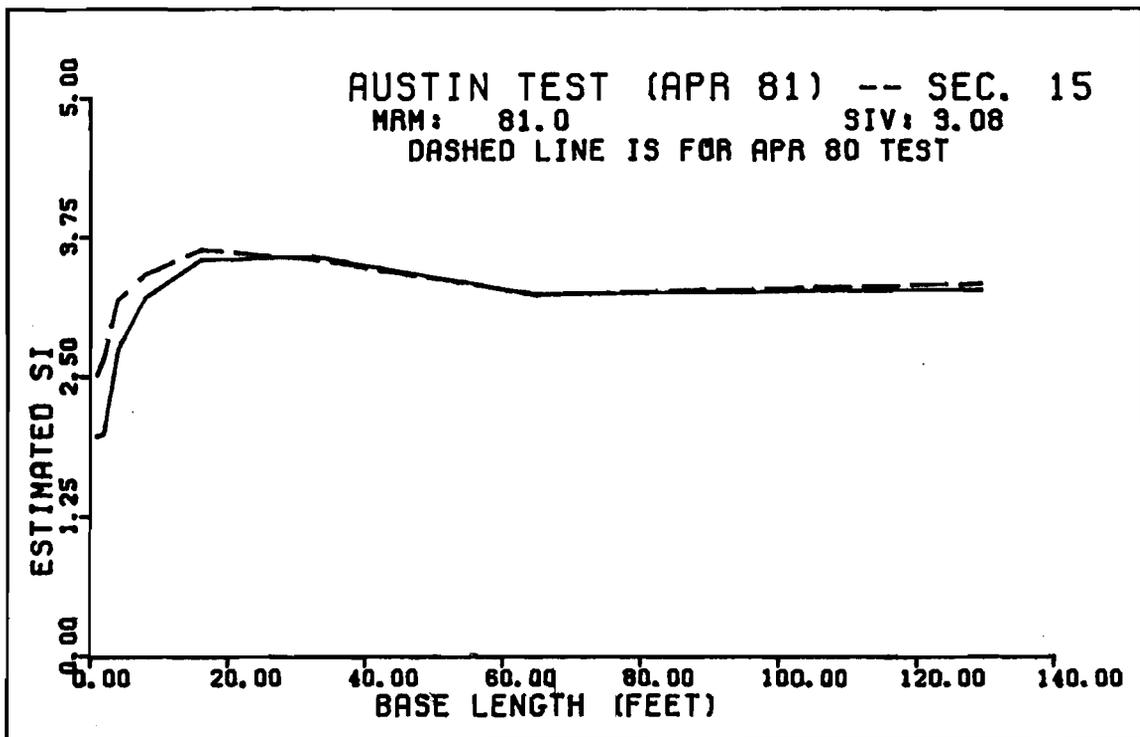


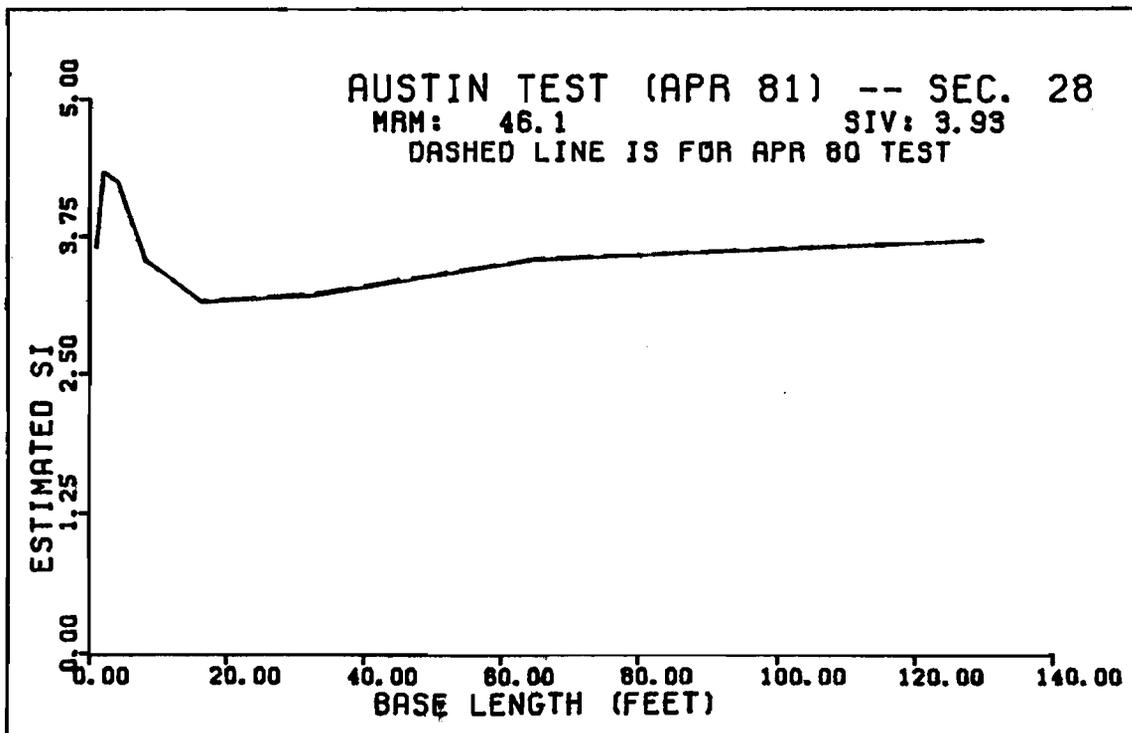
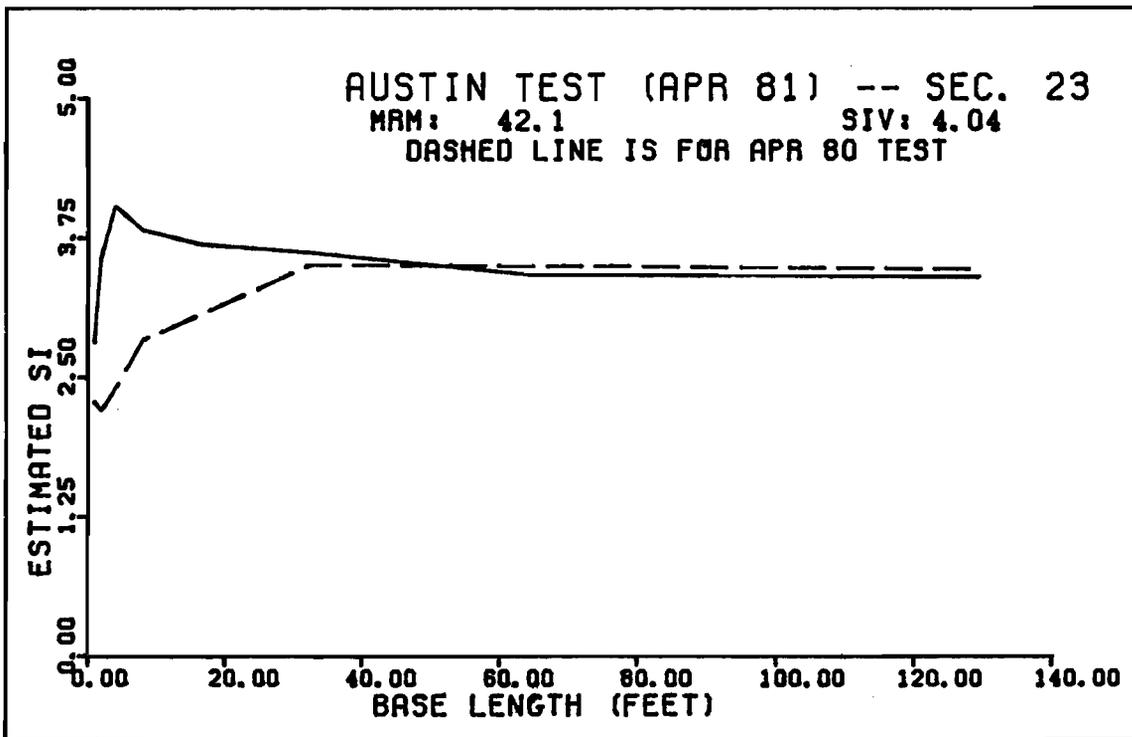


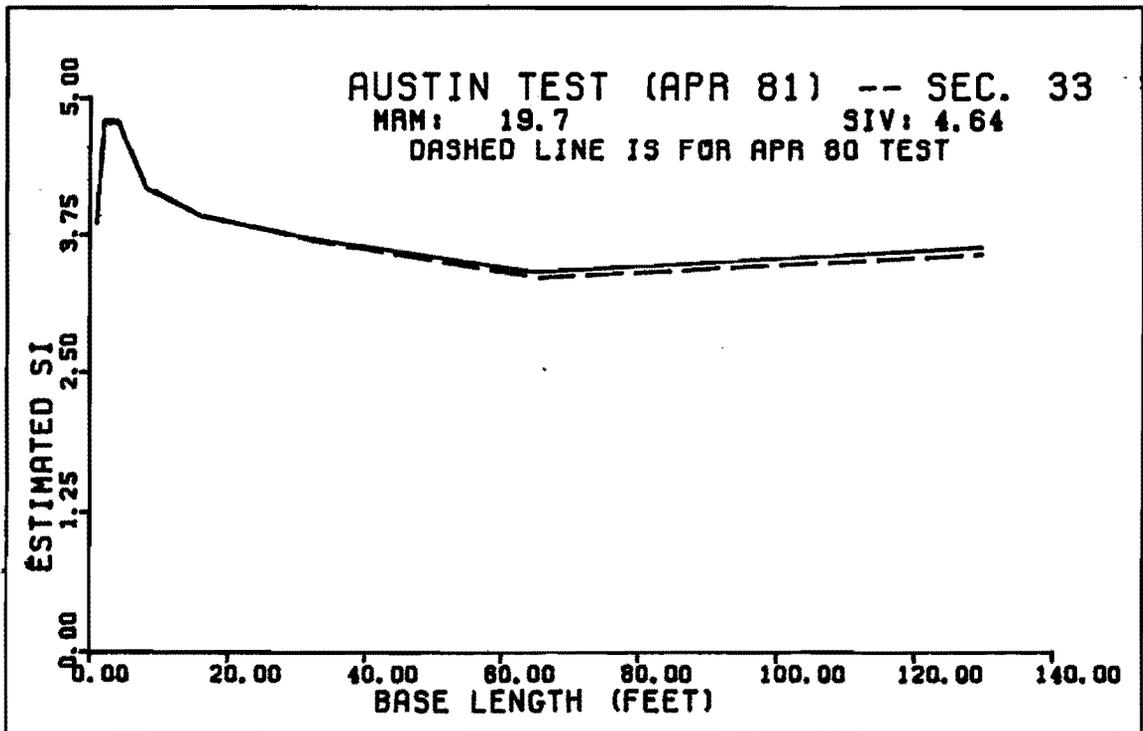
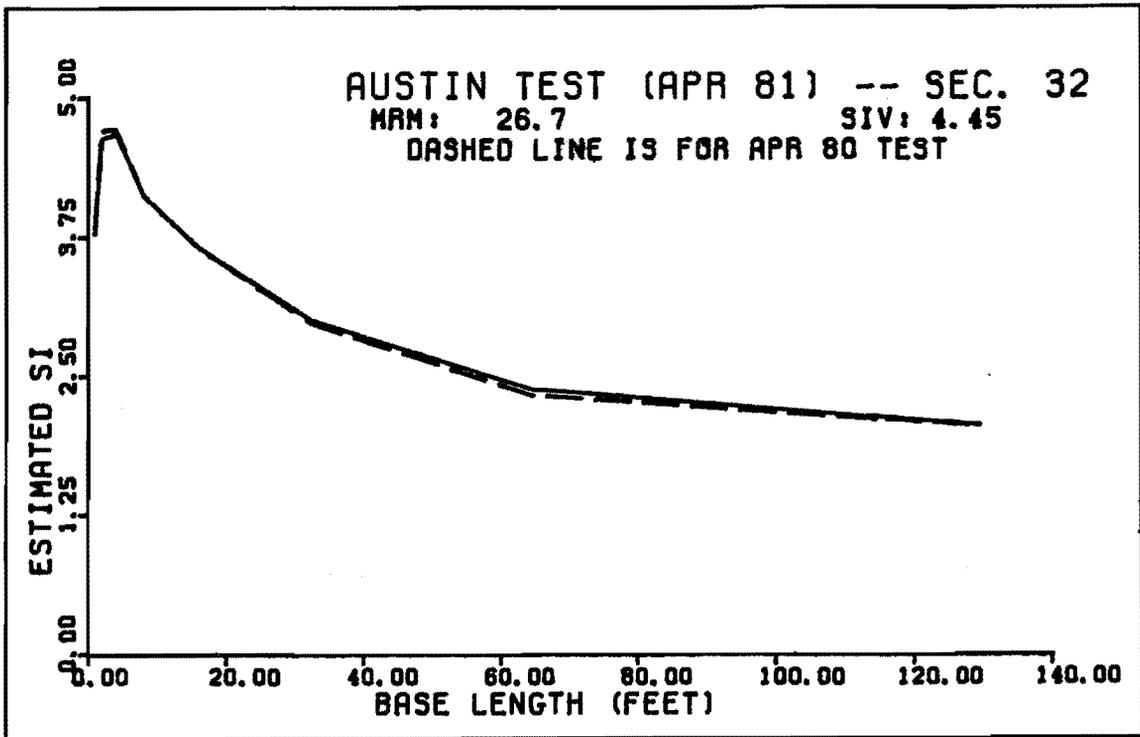


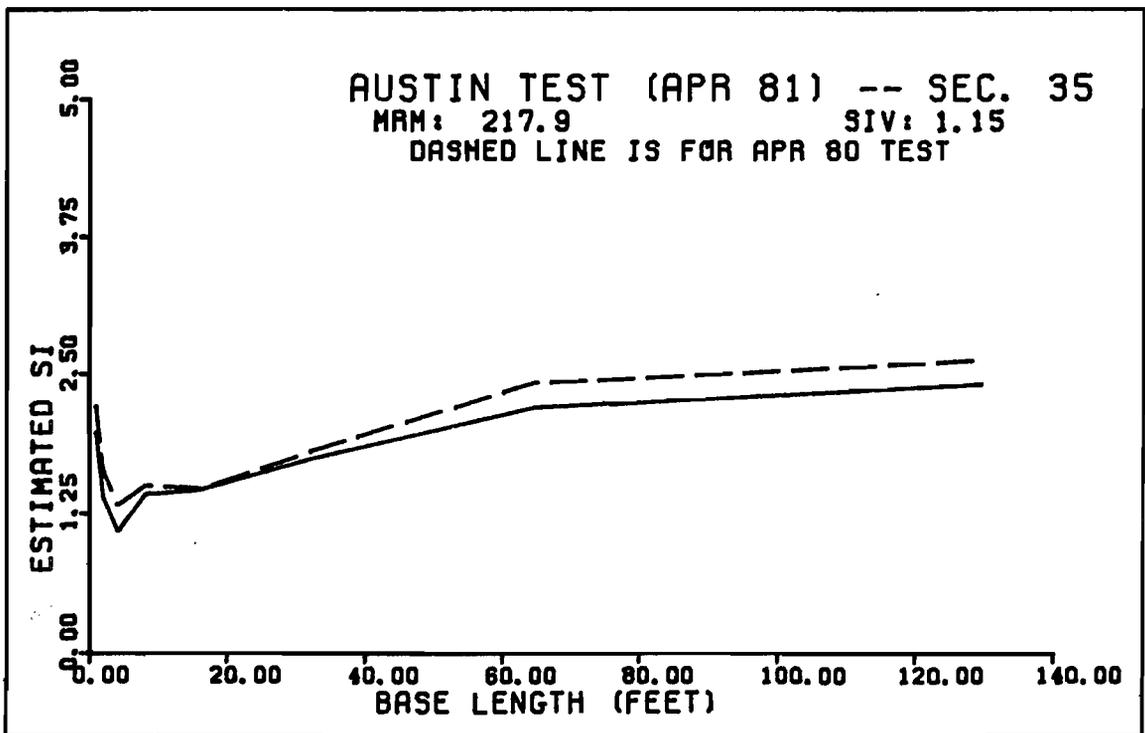
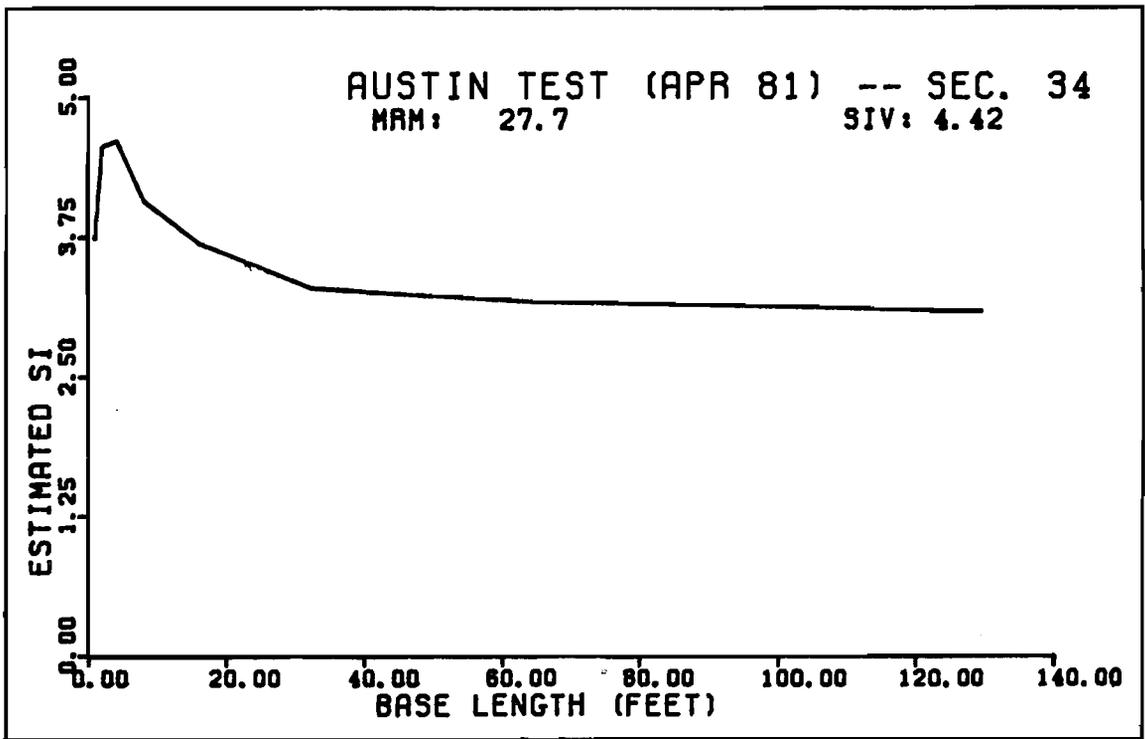


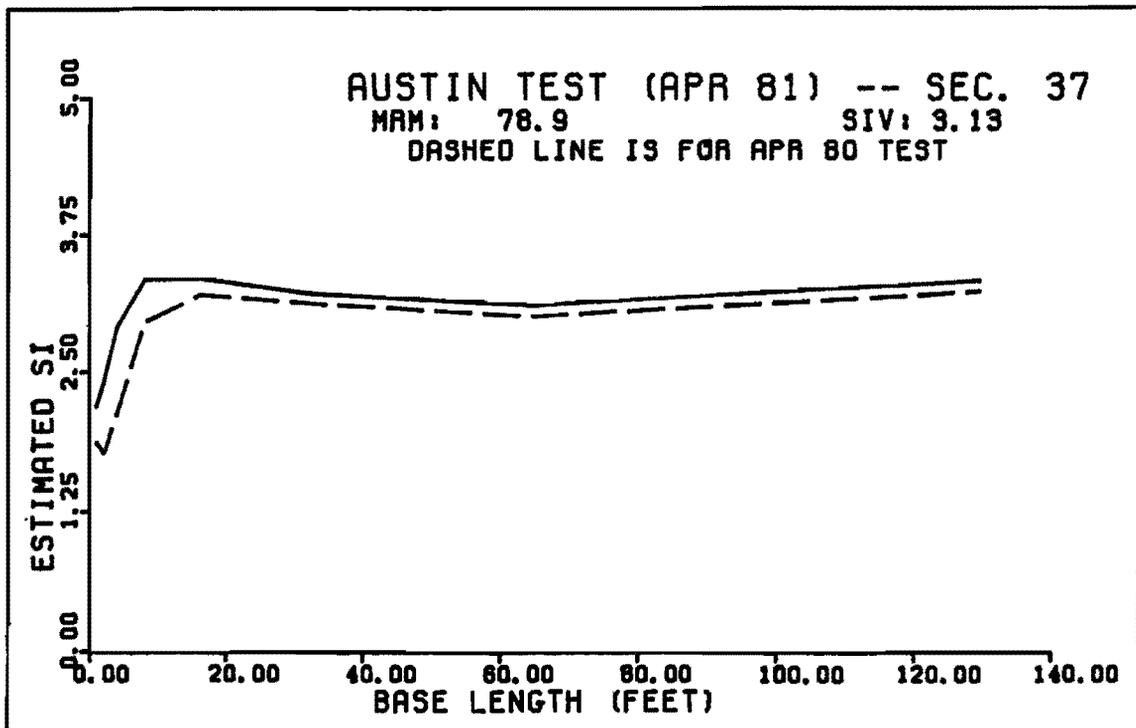
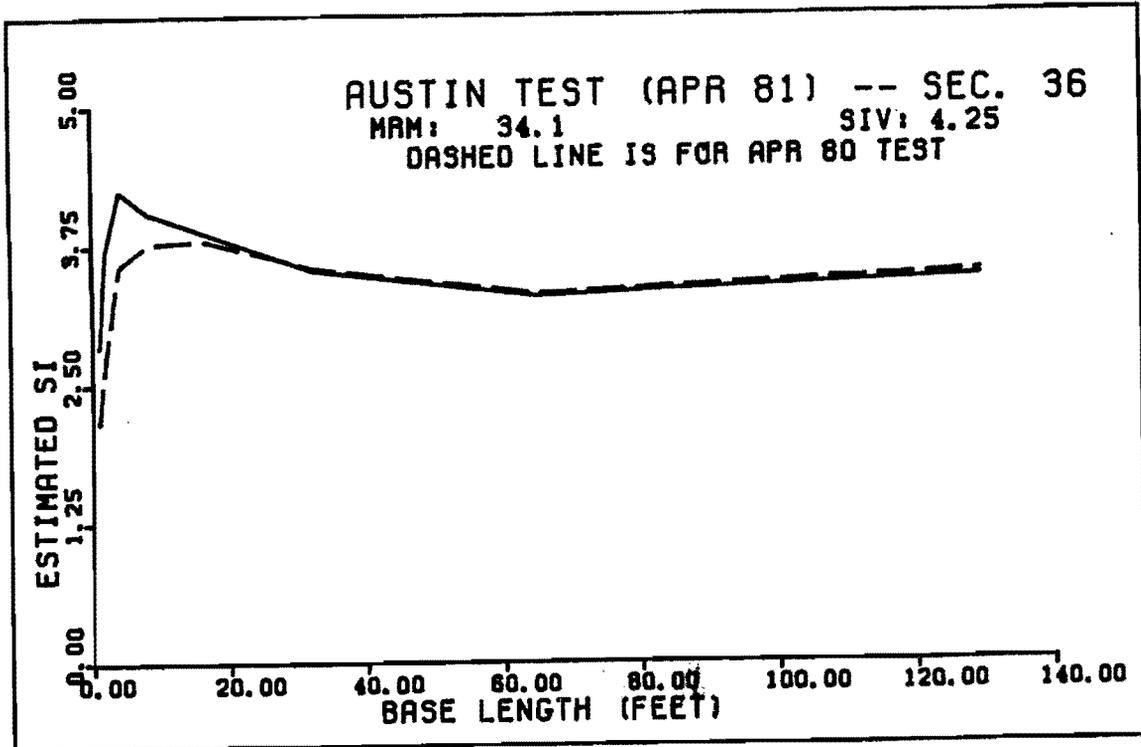


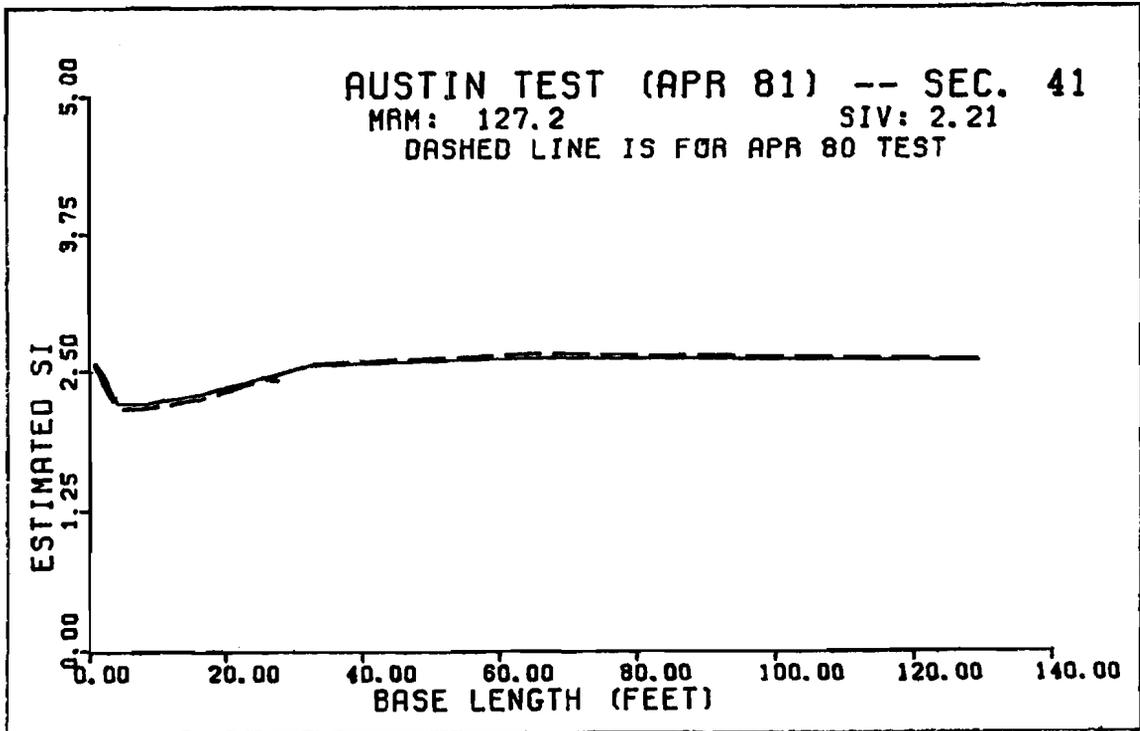
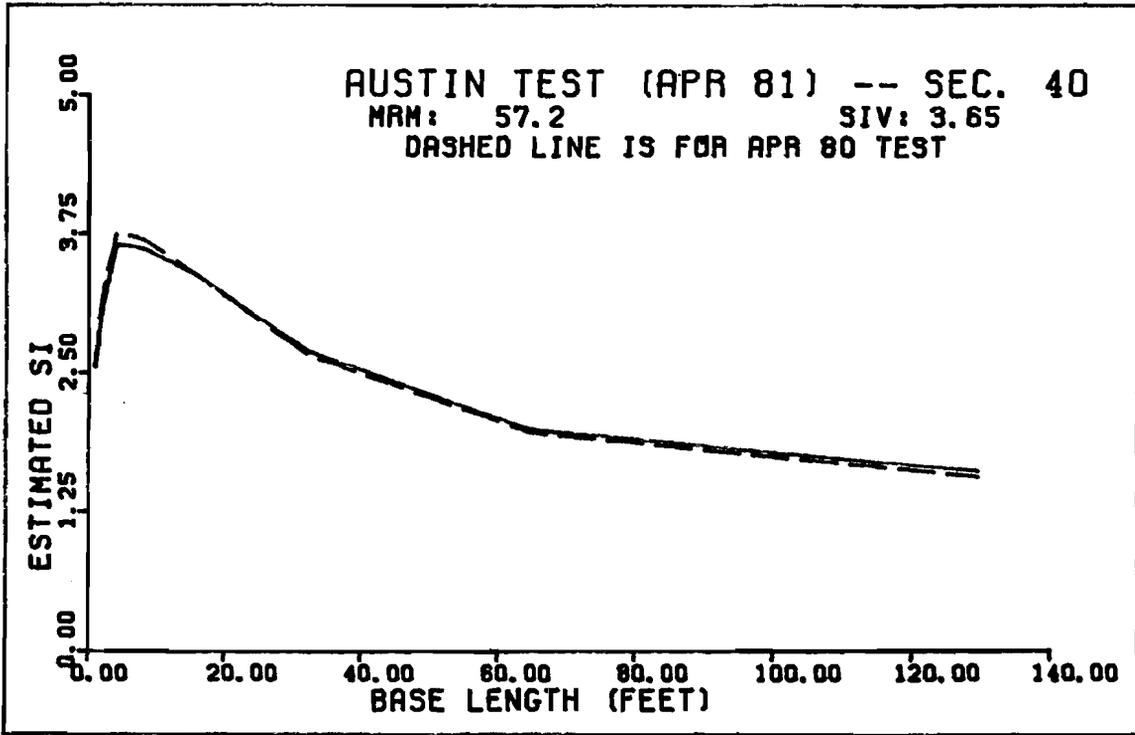


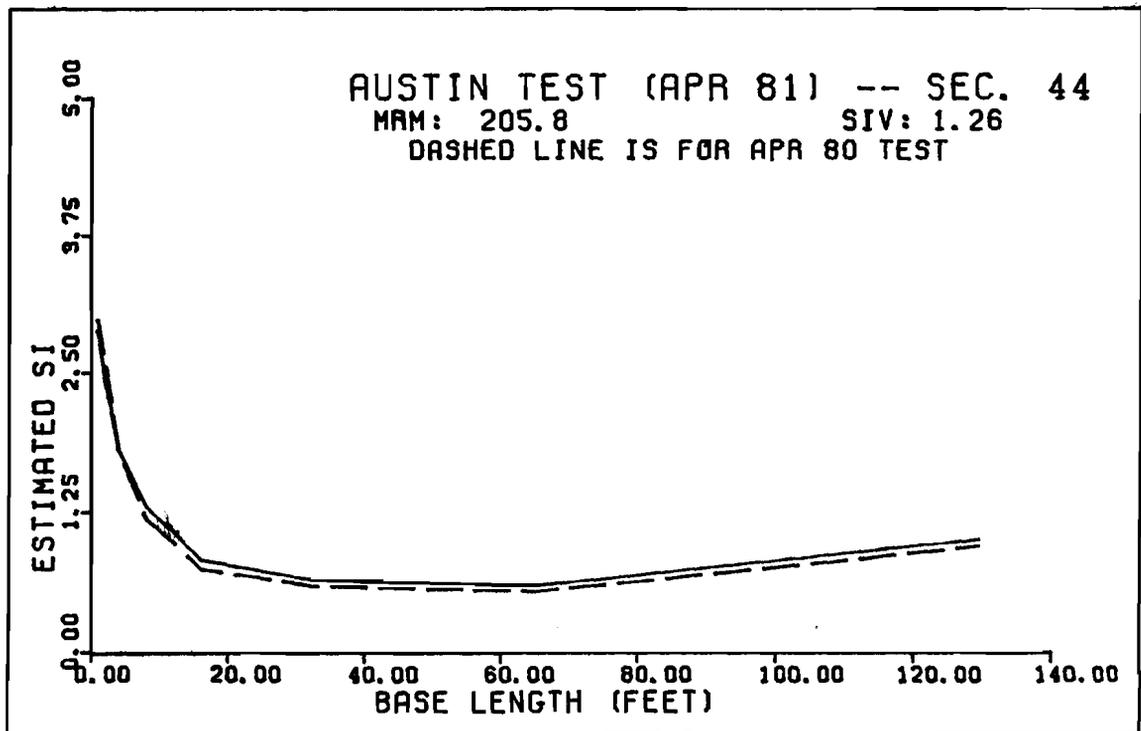
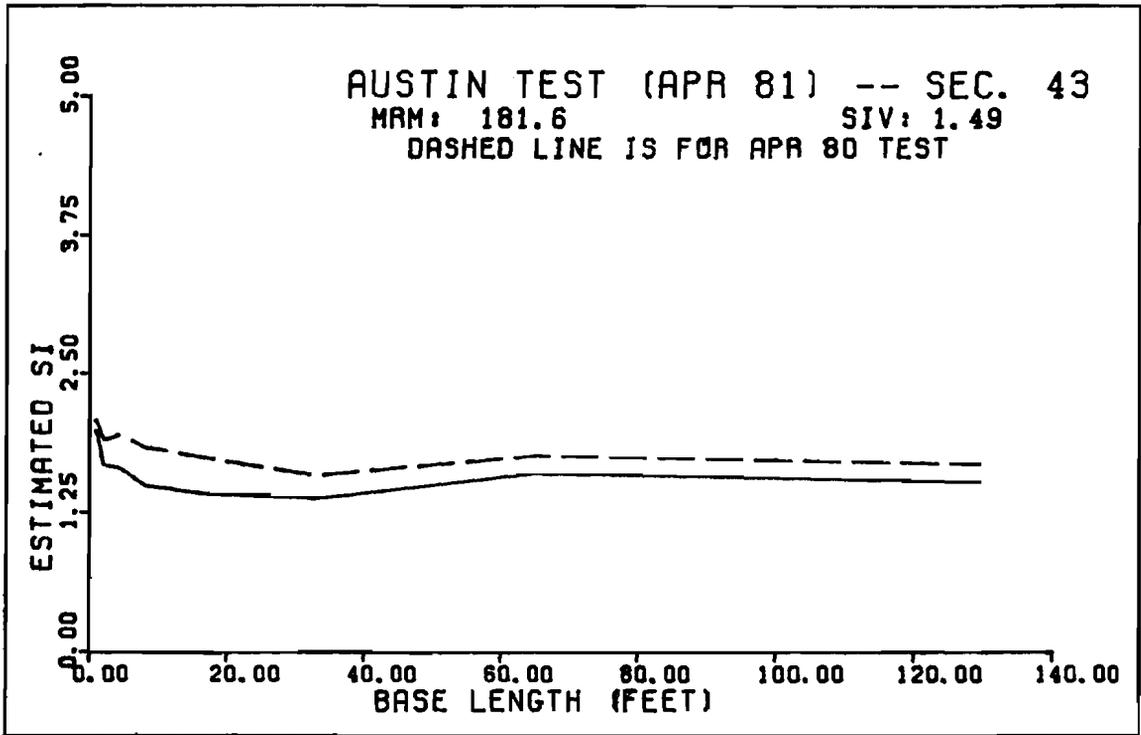


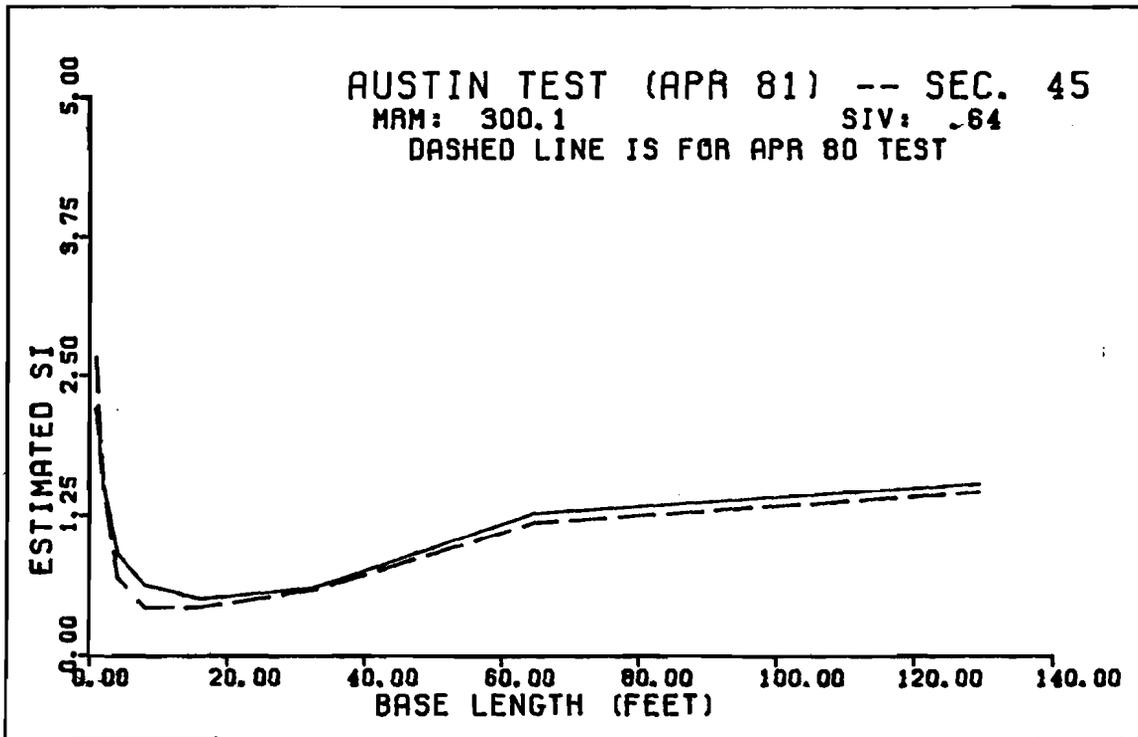












THE AUTHORS

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He specializes in several areas of transportation including weighing vehicles in motion and traffic simulation. He has been active in building the transportation engineering program at the University for the past 20 years and serves as advisor to numerous state and federal agencies, as well as to several industrial corporations.

Presently, he serves as faculty liaison with the International Road Federation and the Transportation Research Board.