

TEXTURE MEASUREMENT AND SKID NUMBER PREDICTION USING LASER DATA
ACQUISITION, DIGITAL SIGNAL PROCESSING, AND NEURAL NETWORKS

By

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ABSTRACT

TEXTURE MEASUREMENT AND SKID NUMBER PREDICTION USING LASER DATA ACQUISITION, DIGITAL SIGNAL PROCESSING, AND NEURAL NETWORKS

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Real-time estimation of the skid number of pavement is difficult. Traditional methods of volumetric measurement are cumbersome and time consuming. It is desired to enable prediction of the skid number of pavement using non-contact means, and to do so using a method which provides a reasonable estimate of the pavements skid number. This research used laser data acquisition of macro-texture, Digital Signal Processing and Neural Networks to estimate the skid number of pavement to a reasonable degree. The research used Digital Signal Processing to identify potentially bad data sets, and a Neural Network model for predicting skid number on the refined data from the DSP. The method enabled relating a statistical index to the texture characteristics of pavements. The model

is based on surface roughness characteristics of pavements as measured by a laser based measurement system, and has the potential to be adapted to a real-time measurement system.

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CHAPTER 1

INTRODUCTION

1.1 Introduction and Summary

This document discusses the estimation and prediction of the Skid Number using linear regression analysis and a neural network, based on data collected with a laser based texture measurement system. Skid Number (SN) is a specific variant of general friction, as pertains to the friction between a rubber tire and a pavement surface. In general, friction is highly non-linear and difficult to predict, model, and analyze [2]*. Skid number prediction, is similarly difficult to predict largely because of the highly non-linear nature of friction.

This research focused on the analysis of the data, and investigated the correlation between the surface texture and the skid number measurement. This research used data previously gathered with a high speed laser based texture measurement experimental system that was previously developed at The University of Texas at Arlington's Transportation Instrumentation Lab.

* All numbers appearing in brackets refer to the numbered references listed on pages 83-85.

The data collected only involved wet pavement measurements, and this research does not discuss dry pavement SN analysis. This research does not cover the development of any measurement hardware, but, for completeness, this document does review the hardware and techniques used to collect the data.

As a first attempt at correlating the surface texture data with measured SN, linear regression methods were used. Correlation of various measurements of the texture with the skid number produced some favorable results. This technique produced R^2 correlation coefficients of about 0.64, which is approximately in line with conventional mechanical volumetric based measurements and predictions.

Second, attempts at analysis were made using Digital Signal Processing (DSP), specifically, the power spectrum. The surface of the pavement was treated similar to a waveform. This technique implemented a neural network in the analysis of the data and skin number prediction. This technique proved inadequate, and R^2 correlation coefficients were only on the order of 0.22. However, one positive result from the DSP procedures shows that the power spectrum analysis of the data can show that some data sets had numerous errors in them and may lead to faulty readings. This DSP data check enables testing if any given data set was actually collected properly, and enables elimination of faulty data. The faulty data was eliminated from future analysis.

Finally, a neural network was developed using the Matlab Neural Network Toolbox, and measurements of micro-texture based on a so-called string filter. The network provided correlation and prediction of skid number from texture measurements with R^2 values of between 0.71 and 0.75, depending on what set of data was analyzed. Portions of these results were presented in [1].

Finally, the development of the analysis techniques showed the potential for improvement in measurement data collection which may lead to improved skid number prediction. Several recommendations, including data on temperature and velocity, are recommended to improve the process.

1.2 Overview

This document discusses the development of the correlation techniques used in predicting SN from micro-texture readings.

Chapter 2 begins with a review of friction as pertains to skid number, macro-texture and micro-texture definitions and measurement techniques. The hardware and file structure are presented for completeness.

Chapter 3 includes the results of linear regression analysis which showed that the correlation was possible. The discussion includes a logarithmic transformation to simplify the analysis.

Chapter 4 includes a discussion of the DSP power spectrum analysis. Although this analysis proved to yield inaccurate prediction, a secondary result enabled finding data sets with large numbers of erroneous data points and elimination of these noisy data sets for future analysis.

Chapter 5 shows the development of a neural network. The inputs to this network were measurements of texture based on a so-called string filter. This correlation and prediction technique yielded significantly better results. These results were presented in [1].

Chapter 6 gives conclusions and recommendations for further research. If future tests include additional data, such as velocity and temperature, the neural network developed in Chapter 5 may yield better results.

The Appendices include various hardware diagrams, alternative, and other data.

CHAPTER 2

BACKGROUND

2.1 Introduction

This chapter provides background information on pavement skid number measurement. The chapter begins by presenting background information on friction and how friction relates to the skid number. Then, the micro-texture and macro-texture of pavement are defined. The chapter concludes by discussing the methods hardware, and file format used for the data collection of the surface measurements and skid numbers. Included is a brief description of the laser system used to collect the data for this research, as well as classical methods. Details of this method can be found in [20]. Appendix A provides more detailed diagrams.

2.2 Friction and Skid Number

Friction is a force which acts to oppose motion. Friction is a dissipative force in that it dissipates energy in a system. Generally, there are two types of friction models, Coulomb, or dry friction, and fluid friction [2]. A general model for dry friction is given by

$$F = \mu_s N \quad (1)$$

where μ_s is the coefficient of static friction and depends on the type of materials in contact. μ_s has been found to be independent of the normal force and the contact area [2]. Rolling friction, sliding friction, and other types of contact friction behave similarly, however, are highly non-linear and may reach a maximum frictional force value.

The skid number, (SN), is a variation on friction specifically between a rubber tire and a pavement surface, and is commonly used at estimating the friction between a vehicles tires and pavement surface. Insufficiently low SN can lead to unacceptably long stopping distances, leading to automobile accidents. SN is defined as

$$SN = 100\mu = 100\left(\frac{F}{W}\right) \quad (2)$$

where W is the dynamic vertical load on the tire and F is the tractive force applied to the tire at the tire-pavement contact and μ is the coefficient of friction [3].

States government agencies generally regulate the minimum acceptable SN for highways, streets, and roads to insure commuter safety. Unfortunately, accurate measurement of SN is a timely and costly process, thus the need for a lower cost measurement of SN.

2.3 Macro-texture and Micro-texture

The surface roughness influences the coefficient of friction. A very smooth surface generally has a lower coefficient of friction than does a somewhat rough one. There are two divisions of surface roughness used to describe pavement: macro-texture, and micro-texture. The skid resistance, or frictional properties, of a pavement surface in contact with

a given tire under similar conditions has been shown to be largely determined by the texture of the surface [3-6].

Macro-texture is defined as the deviations of a pavement surface from a true planar surface with the characteristic dimensions of wavelength and amplitude from 0.5 mm (0.02 in.) up to those that no longer affect tire-pavement interaction [3]. It is generally believed that macro-texture provides the drainage on wet pavement and determines the rate that skid resistance decreases with increasing speed. Pavements with more macro-texture typically show a smaller rate of decline of skid resistance with increasing speed.

Similarly, surface micro-texture is defined as the deviations of a pavement surface from a true planar surface with characteristic dimensions of wavelength and amplitude less than 0.02 inch (0.5 mm) [3]. The sharp, fine particles in a surface which make up the micro-texture penetrate the thin water film and permit contact of the tire and the pavement surface. Many consider the micro-texture to be the predominant factor in determining the skid resistance of pavement at low vehicle speed. Macro-texture is somewhat easier to measure because of the greater size.

It appears that both macro-texture and micro-texture should be considered when determining the skid resistance of a pavement. Measurements of macro texture have been reported using optical techniques, pavement profiles, and the sand-patch method [7, 19]. Micro-texture measurements are often reported using the British Pendulum Number [3, 4]. Researchers have reported results obtained from profiles using optical techniques or a mechanical stylus, but these profiles are difficult to obtain [20]. Lasers provide very accurate distance measurements and are now being used for many such measurements. A complete compilation of all the many studies for macro and micro texture are not

included herein. A good report of much of the early work, as well as recent work, is provided in [20].

2.4 Classical Mechanical Texture Measurement

Classically, surface texture has been measured using the so-called sand-patch method discussed in [19]. This method uses a volumetric approach to estimate the surface texture, however, the granule size limits the accuracy, and the test is generally used to estimate macro-texture, as opposed to micro-texture. Because the test involves spreading material similar to sand on pavement and measuring the size of circle when smooth, the test is cumbersome and generally the standard deviation is about 27% of the estimate of the actual macro-texture [20, 21]. Correlating this surface texture with SN is difficult. Further, the test is generally inadequate for measuring micro-texture.

2.5 Data Collection and Preliminary Tests

The actual data used in this study were collected using a laser based scanning system developed at The University of Texas at Arlington's Transportation Instrumentation Lab in a previous study [20]. Two different systems were developed, one for collecting micro-texture, and the other for macro-texture. Background information on the equipment used is discussed below. The macro-texture data was collected at various sites in Texas in February and June of 2000. These sets of data will be referred to hereafter as the "02" and "06" data.

A high resolution laser scanning system was developed to obtain accurate measurement of the micro-texture, but could not be used at high speeds. Thus only

macro-texture was measured for the high speed 02 and 06 data. The next section will discuss the micro-texture system followed by a description of the high speed macro-texture measurement system.

Prior to collecting micro-texture data on pavement, a simple laboratory test was conducted to test the resolution by scanning four different types of sandpaper with grits of 36, 50, 80, and 120. The results show that the different sandpaper grits could be clearly differentiated and measured. The laser can measure up to 0.2 micrometers.

During preliminary testing, office tile joints were also tested. Figures 2.1, 2.2, and 2.3 show the results of these tests.

In preparation for collecting the data, a representative 15 foot (4.6m) section of the particular pavement type to be measured was selected. A trace was marked to obtain repeat runs and 2 markers were placed on the path at approximately 10 and 11 feet (3 and 3.4m) to validate the collection speed. Details of these results are given in [20].

2.6 Measurement Hardware and Embedded Systems

This section reviews the laser scanning system hardware and software used to collect the data. A complete parts list, operation procedure and parts specification is given in [21].

The prototype scanner system included two 150mm travel translation tables with a precision distance measurement laser. The translation tables have a one micrometer step size and at a maximum speed of 2500 micrometers per second. The laser measured distances within a ± 3 mm distance from a 30 mm standoff from the target with a 0.2 micrometer resolution. The laser spot size varies from (0.05 by 0.12 mm at -3 mm), (0.07

by 0.03 mm at 0 mm), and (0.15 by 0.07 mm at +3 mm) from the 30 mm standoff distance.



Figure 2.1: Office tile joint used in test.

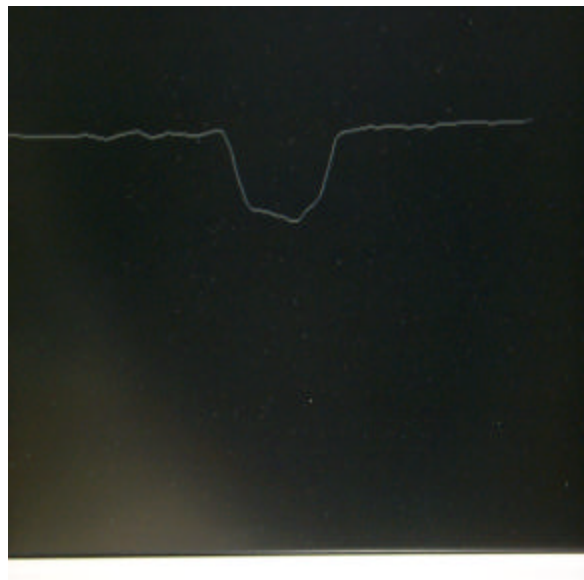


Figure 2.2: Texture Scan of two joints.

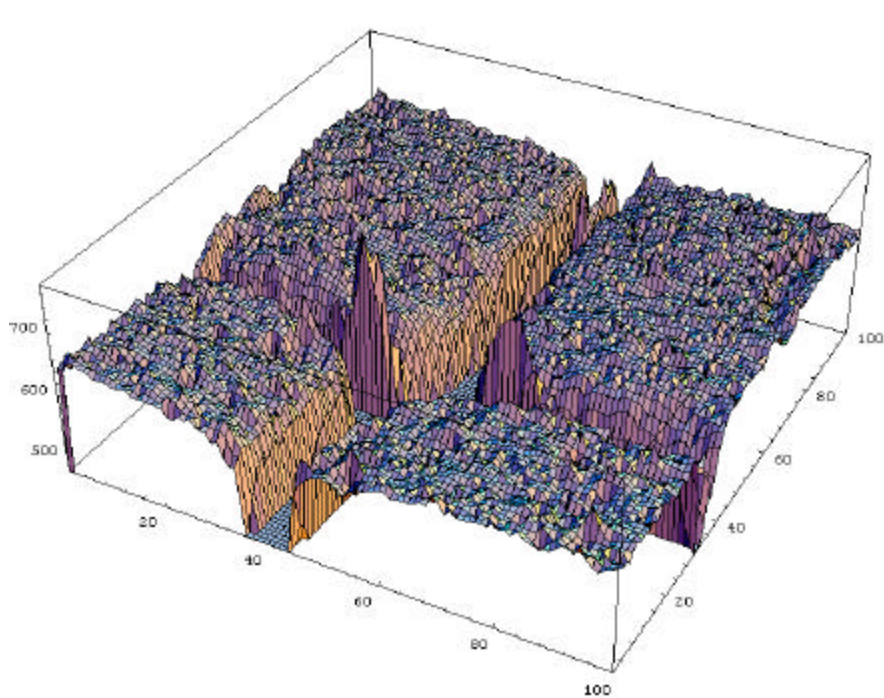


Figure 2.3: Scan of office tile joint.

The user specifies the number of X and Y samples and the sample spacing to the scanner program. The software scans the target in the X direction at the various Y positions, and samples the laser output at one micrometer steps in the X scan direction and stores the average sample distance within the specified X sample spacing. The results are stored as an ASCII file on disk.

The Scanner Controller consists of a HC11 microprocessor, a 16-bit A/D converter, and various circuits for synchronization with the MM2000 Universal Interface Box and a LPT port of the PC. The controller used a Motorola M68HC11EVBU Universal Evaluation Board as the micro-processor system for the controller operation, which provides the A/D conversion of the LM200 analog signal. The system places a 32-bit

integer representation of the analog signal into a FIFO, which can be read, with the LPT port of the PC. The LPT port is used for sending values and commands to the controller.

A bi-directional LPT port is not required to interface with the Scanner Controller.

A few minor modifications were made to the BD-F15 connector to provide X-axis encoder signal to the A/D Acquisition and Laser Interface Controller.

2.7 A High-Speed Texture Measurement System

The primary sensor used by the system for the high speed macro-texture measurements is the Selcom texture laser. A diagram of the components in this system is illustrated in Figure 2.4. Currently, Selcom lasers are used by the majority of road profiling measurement systems both in the United States and internationally. Selcom also offers a high resolution laser that is now being used for high speed texture measurements on other systems. The Selcom lasers have proven to be very reliable and effective devices for displacement measurements in the road measurement environment. Efforts were made in the early part of the project to find suitable candidates for texture measurements, but none were found to be significantly better than this laser. Because of the lack of other candidates systems, the pilot study described earlier, and the fact that the Selcom texture lasers are used elsewhere for texture measurement the Selcom laser 78KHz -16 micrometer laser was selected. This section discusses the high-speed texture measurement system developed for the project. The discussions will include the system architecture, followed by details on the various modules used for texture measurements. The characteristics of the sensor data are also described. Finally, the laser interface

system is discussed. The next chapter describes the program and procedures that were developed for collecting texture data.

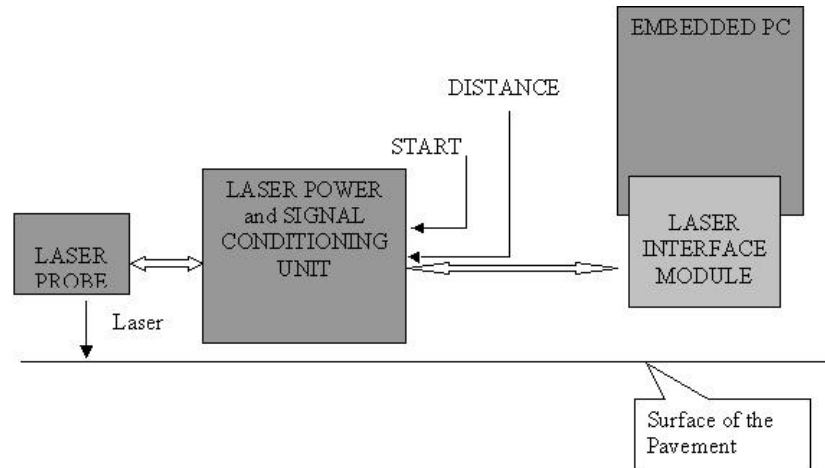


Figure 2.4: Diagram of components.

As indicated, the laser, distance and start signals are interfaced to the PC via the PC Laser Interface Module or PCLIM. During the first year of the project, efforts were made to locate appropriate data acquisition equipment for both the micro-texture and macro-texture measurements. Selcom offered a VME based system board for their lasers, where the laser displacement readings could be read either as an analog or digital displacement reading. The digital values could be obtained directly in digital form via the VME bus, or the displacement values could be read directly from the laser optocoupler as a synchronous serial data stream. Initial efforts were directed toward reading the signal in analog form, and then digitizing the results by the data acquisition system so they could be recorded or processed. At that time, most profilers and texture measuring systems obtain laser data

for the various surface measurements in a similar manner. After investigating the analog and averaging method used by Selcom for providing the analog laser readings, and because of the VME board requirements, it was decided to build a special board for this application. where the laser displacement readings were obtained from the serial data stream and converted to a parallel value for direct input to the PC via the ISA or PC104 bus. The requirements for the board design included a direct interface between the serial data stream sent by the laser probe and the PC via the system bus. Thus, the method would provide a reduced noise advantage in that noise or errors introduced by taking the digital displacement readings, converting them to an analog form, and then once again back to digital would be avoided.

The Selcom system provided a selectable running average on a special averaging board that provided averaged displacement readings. It was decided to include the averaging function and allow the software to specify the amount of averaging. As will be later discussed, averaging the continuous laser readings is common and usually necessary in most laser systems. The Selcom averaging board discards readings that are detected as bad from the laser probe. That is, during measurements, if one of the continuous laser readings were found bad (a laser return signal was not received), it would be discarded and replaced by the previous reading (assuming it was good). If readings are discarded, then the number of time-readings between two consecutive distance-readings could vary for the same speed, thus making it impossible to obtain accurate longitudinal wavelength information. The above considerations became requirements for the PCLIM board. The system would let the software determine how to average the laser readings, as well as maintaining wavelength integrity. The PCLIM provides a way to read the laser values

directly into memory via the ISA or PC104 data bus (via an ISA to PC104 adapter). The next section provides details on the architecture of the Laser Power and Signal Interface Unit. Figures 2.5 and 2.6 shows the laser components and bus, and Figure 2.7 shows the laser installed on the trailer used in the data acquisition. Several tests were made at various test sites in Texas.



Figure 2.5: The Selcom laser system.

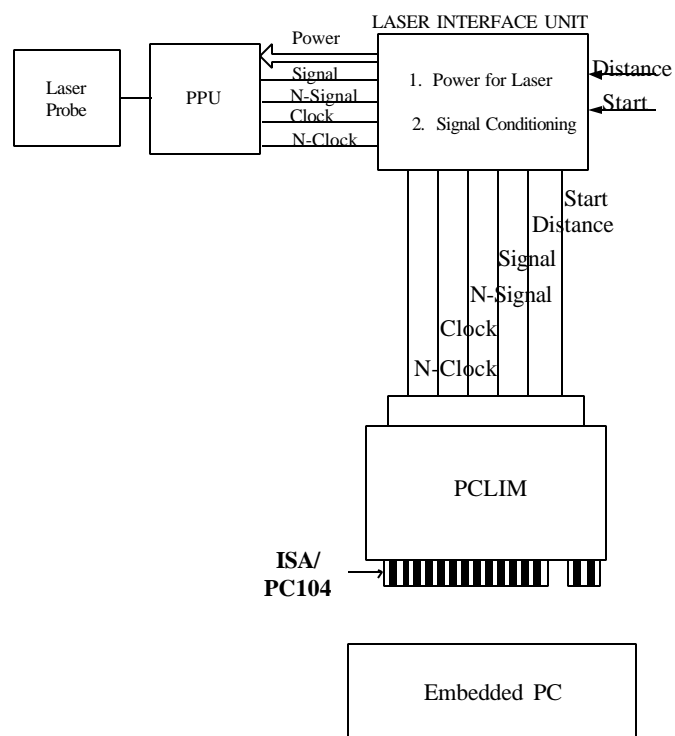


Figure 2.6: Bus schematic.



Figure 2.7: TxDOT trailer with laser.

2.8 Data File Format

Each section of pavement scanned was stored in file of data referred to as a data set. The raw test data was stored in files of the format shown in Figure 2.8. Files were stored in the format MMDDHHmm.SS where MM is the month, DD is the day, HH is the hour, mm is the minute, and .SS is the second. For example file 08031433.19 was collected on August 3rd at 14:33:19. These files contained the raw profile data read by the laser along with header information. The header information includes various parameters governing the laser setup and other information. The first four bits of each data point were a set of flags set when the point is taken. These flags indicate if there is something wrong with the specific data point. These points were later analyzed by the string program, discussed in the next section. Bad points were ignored. However, as will be discussed in Chapter IV, sometimes entire data sets (files) were also bad. These data sets contained so many faulty readings that in some cases the data processing programs could not even process the data, and in other cases, the power spectrum was significantly altered. If significant faulty data was found to be in the file, to the point where calculations of MPD and other measurements could be in significant error, the file was discarded. This process will be discussed in Chapter 4. Individual bad points in a data set were removed when analyzed by checking the flags in the first four bits.

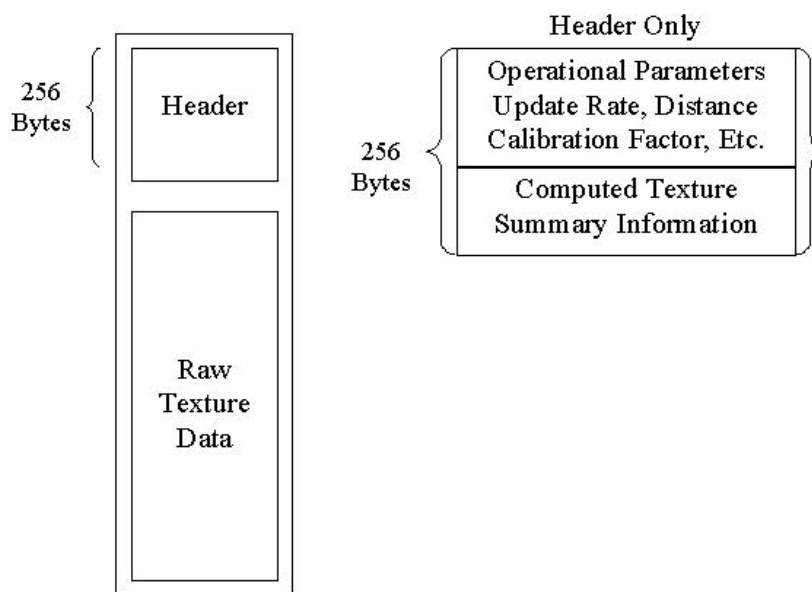


Figure 2.8: Data file format.

2.9 Texture Variables and String Filtering

A statistic frequently used in describing texture is mean profile depth or MPD. The computation of this statistic has been well defined both by the ASTM and the ISO. The two dimensional surface profile for a small section of a pavement is illustrated in Figure 2.9. The profile is given in terms of amplitude and distance. Inertial reference profilers provide an estimate of the surface profile with long wavelength information removed. An estimate of the true profile can be measured by rod and level. There is considerable information defining profile frequency or wavelength, amplitude, as well as the measurements methods of road profilers and thus these concepts will not be repeated here. Texture Profilers provide measurements of the very short wavelength, related to the definitions in section 1 for micro and macro-texture. These systems must use a horizontal (sampling frequency) and vertical displacement measuring method such that the definitions of the bandwidth ranges for micro and macro-texture are met.

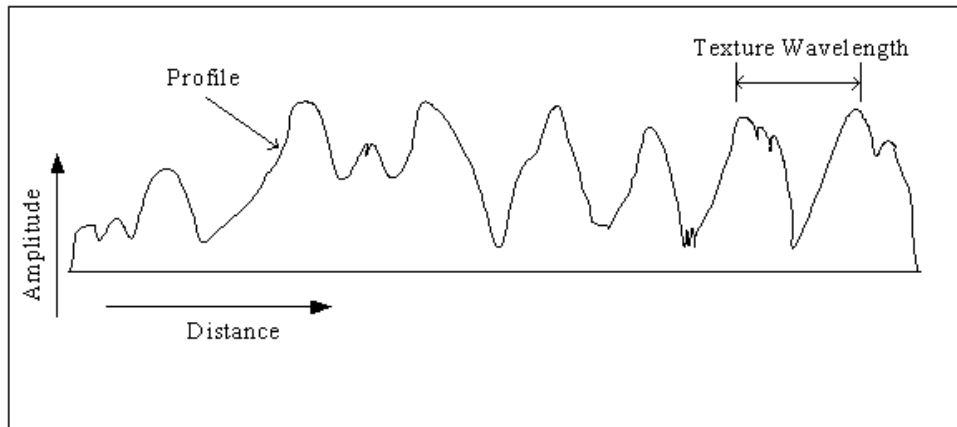


Figure 2.9: Diagram of pavement surface.

For longer wavelengths such as 0.5 feet to 500 feet, the combination of a displacement measuring laser and accelerometer are used to remove the influence of the vehicle suspension system in Road Profilers. For the much shorter wavelength range of surface texture, the vehicle influence is simply filtered out.

Figure 10 shows the Mean Profile Depth statistic. This statistic is simply the average value of the profile depth over a specified distance (window or referred to as the baseline). Estimated Texture Depth (ETD) is used to estimate mean texture depth and is defined by the equation:

$$ETD = 0.2 + 0.8 * MPD \quad (3)$$

The baseline is typically 100 mm or 3.94 inches (related to the diameter of the spreader tool used in the volumetric method). It would seem that ETD is a primarily a measure of macro-texture and not micro-texture, because of the much greater amplitude

range of the longer wavelengths, although the effects of micro-texture could perhaps be included if the horizontal sampling is sufficient.

After the profile data was read in from the tests, investigations to see if the skid number could be predicted using this profile data were conducted. The data was first pre-processed using various methods. One method found to be useful is based on the so-called sand patch method discussed in [3]. Here the surface texture is related to the diameter of a specified volume of sand-like substance distributed over the pavement. Although somewhat crude, this method yields surprisingly good results.

Similarly, as developed in [11], a string-filter method used for computing rut is a similar method of modeling the time-surface interaction. The string filter is applied as illustrated in Figure 2.11. The filter requires two parameters, base length or BL, and string length, SL. The filter is somewhat similar to the European String Method used for rut measurement, in which an imaginary string is stretched across a set of points, P_1 to P_{16} as indicated in Figure 2.11. The set of points between SL are replaced with a new set, where all points making contact with the string have a zero value and those that fall below the string are replaced with the difference between the point and the imaginary string. The new set of 'string filtered' points are then further divided into one or more subsets, where the length of the subset is BL. In Figure 2.11, the set of filtered SL points consists of 3 subsets, BL_1 , BL_2 , and BL_3 . A second variation of the string filter consists of including in each subset BL_i , only those points that do not make contact with the string. Although not necessary for use with each statistic, one or more of the above statistics can be computed for each BL subset and an average of the statistics for the SL

set determined. For example, using the RMS statistic (also referred to as STD) and using the second variation we get for the first two subsets:

$$SL = (P_1, P_2, P_3, \dots, P_{14}, P_{15}) \quad (4)$$

$$BL_1 = (P_1, P_2, P_3, P_4, P_5, P_6) \quad (5)$$

$$BL_2 = (P_4, P_5, P_6, P_7), \text{ etc.} \quad (6)$$

$$STD_1 = \frac{\left[\sum_{i=1}^3 a_i^2 - \frac{\left(\sum_{i=1}^3 a_i \right)^2}{3} \right]^{1/2}}{\sqrt{2}} \quad (7)$$

$$STD_2 = \frac{\left[\sum_{i=4}^7 a_i^2 - \frac{\left(\sum_{i=4}^7 a_i \right)^2}{4} \right]^{1/2}}{\sqrt{3}} \quad (8)$$

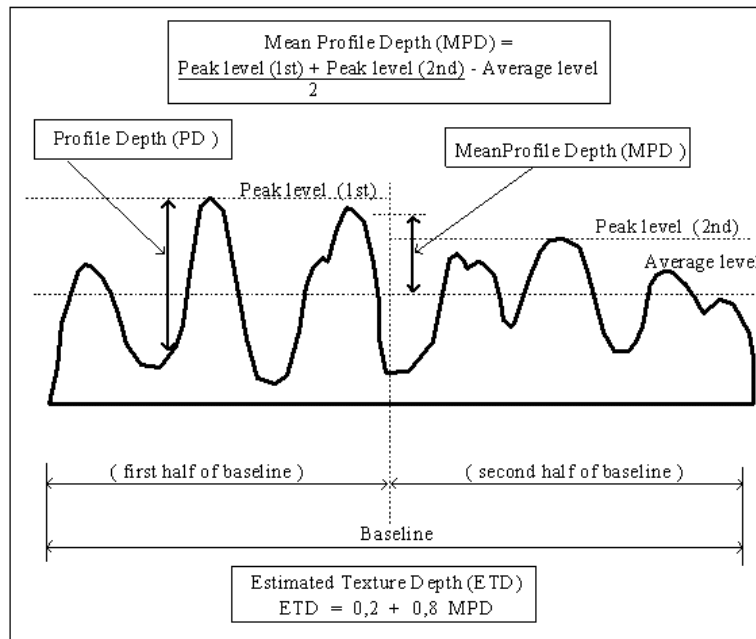


Figure 2.10: Definition of terms.

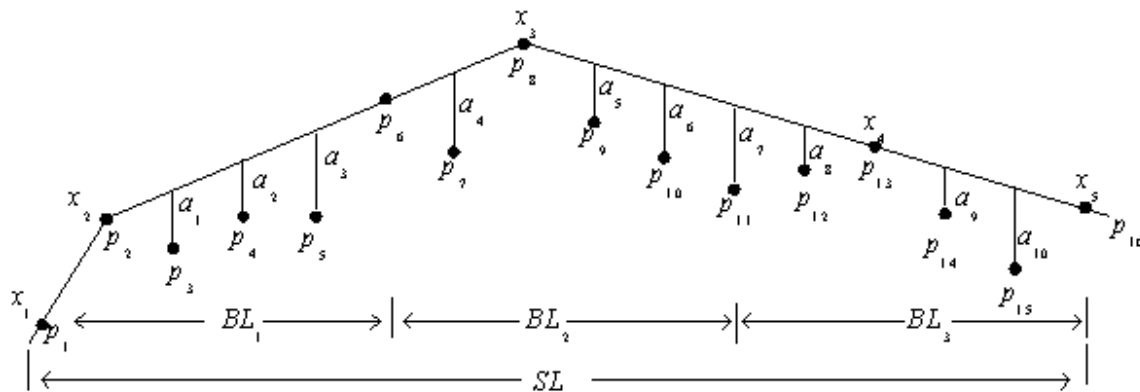


Figure 2.11: Diagram of the string filter.

CHAPTER 3

INITIAL MODELING EFFORTS

3.1 Introduction

Linear regression methods were used as a first attempt to correlate parameters used to measure surface texture and SN. This chapter reviews the results of the linear regression analysis. It was found that the STD discussed in Chapter 2 covering the entire data set (overall STD) produced reasonable correlation with the measured skid number after the data was transformed using a logarithmic transformation. The regression and transformation techniques used are discussed in [12, 13, and 14].

One of the characteristic of this method is that only one texture measurement technique is correlated with the measured SN. Combining numerous variables will be discussed in Chapter 5.

3.2 Initial Analysis

Various forms of texture measurement, such as MPD, discussed in Chapter II were run on the 02 data, to see if any correlation with SN could be achieved. All of these

statistics were based on the string filter using various base lengths and string lengths. Through experimentation, it was found that a base length of 10 and a string length of 150 produced some form of correlation in the form of a logarithmic curve. Figure 3.1 shows one of the logarithmic type curves generated by using the overall STD for string sections and the given correlation with the measured SN from the 02 data. Similar results were found for MPD, ETD, and Average STD. The R^2 correlation coefficient of 0.61 is marginally acceptable, but there is some correlation with the SN. The R^2 correlation coefficient is defined as the square of r given by

$$r = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sqrt{\sum (x - \bar{x})^2 \sum (y - \bar{y})^2}} \quad (9)$$

Because of this initial hint at correlation, a logarithmic transformation was implemented to attempt improvement of the data with a linear correlation. The derivation of this transformation and the results are presented below.

3.3 Logarithmic Transformation and Data Correlation

To facilitate the data analysis, a logarithmic transformation was developed and is derived below. For this model, a general inverse exponential curve given by

$$y = a - \frac{d}{x^b e^{c-d}} \quad (10)$$

Rearranging gives

$$a - y = \frac{d}{x^b e^{c-d}} \quad (11)$$

Taking logarithms gives

$$\ln(a - y) = -b \ln x + (\ln d - c + d) \quad (12)$$

Through experimentation, it was found that letting $a = 60$, $y = SN$, $-b = a$, $x = STD$, and $\ln d - c + b = B$, the following transformation was generated.

$$\log(60 - SN) = a \log(STD) + B \quad (13)$$

It was found that $a=1$ and $B=0$ enabled better correlation. Table 3.1 lists some of the data used after the transformation and Figure 3.5 shows a plot of the data with correlation. The overall correlation increased to 0.64. However, further improvements are desired. Similar plots were made for MPD, ETD, and Average STD, and these results are presented in Figures 3.6, 3.7, and 3.8. A similar transformation as above was used to generate the data for these plots.

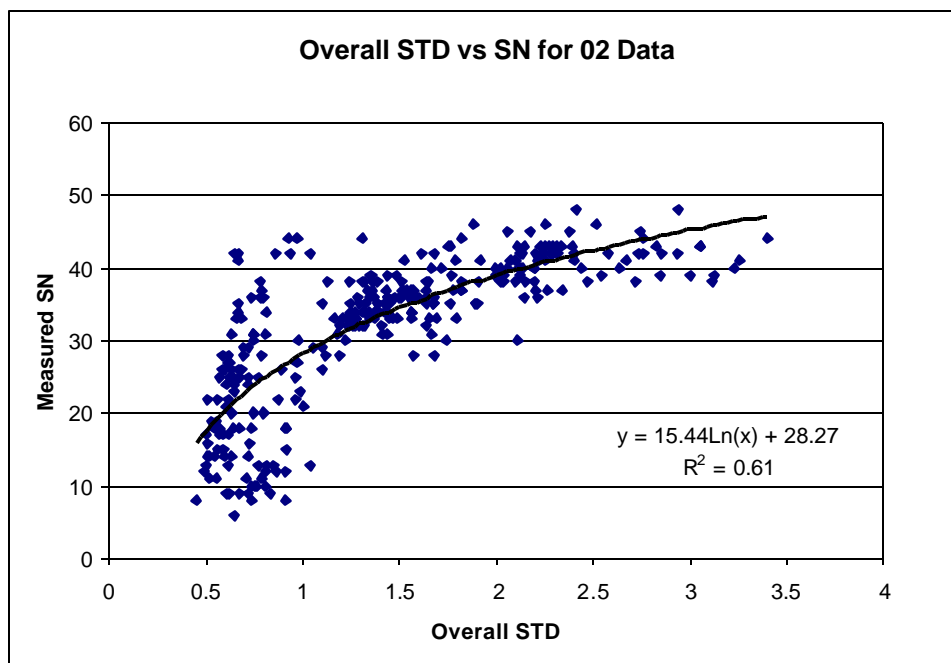


Figure 3.1: Initial Correlation between Overall STD and SN.

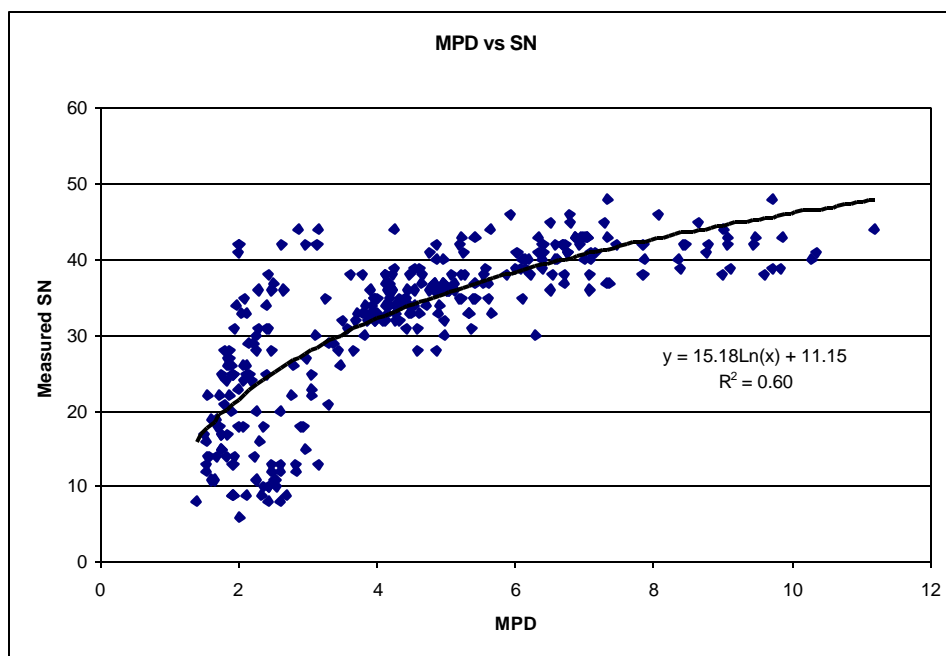


Figure 3.2: Initial correlation between MPD and SN.

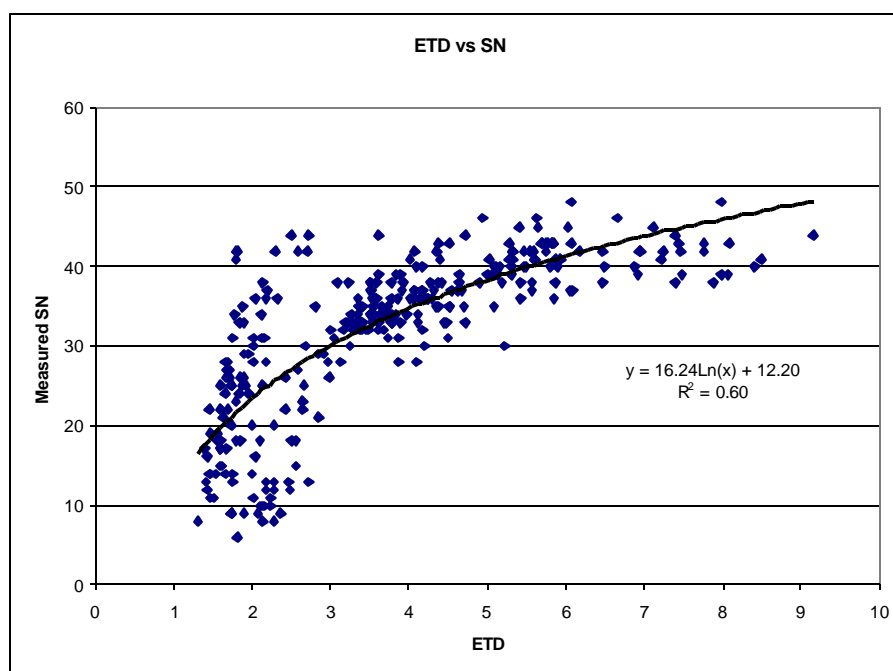


Figure 3.3: Initial correlation between ETD and SN.

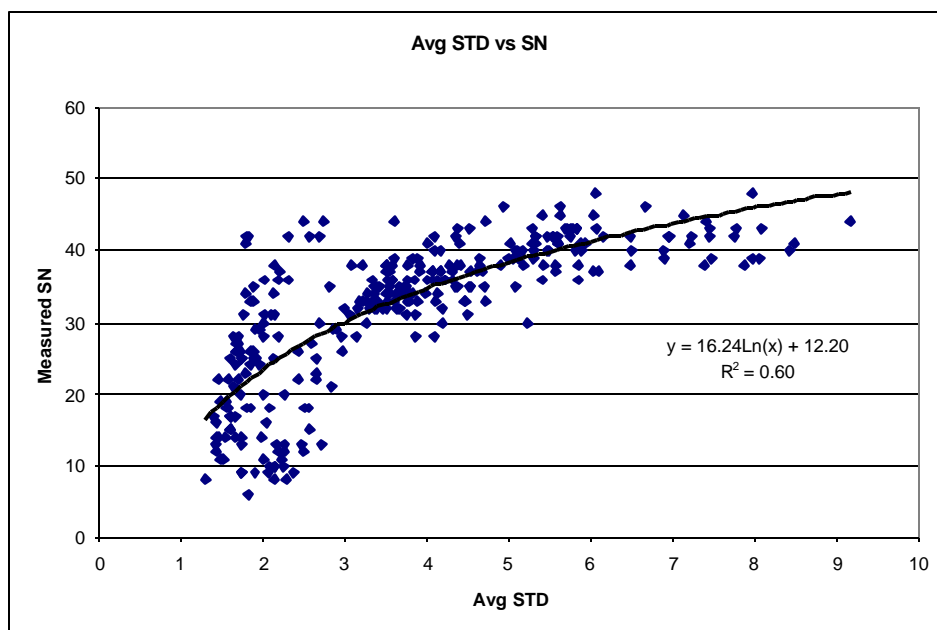


Figure 3.4: Initial correlation between Average STD and SN.

Table 3.1: Example Samples.

Filename	MPD	ETD	Av. STD	Overall STD	Log(STD)	SN	Log(60-SN)
2091052.47	5.407	4.525	1.612	1.889	0.2768	35	1.398
2091052.55	3.445	2.956	1.038	1.122	0.050	28	1.505
2091053.01	4.979	4.183	1.503	1.638	0.2147	32	1.447
2091053.08	6.704	5.563	1.978	2.202	0.343	37	1.362
2091053.15	7.839	6.471	2.261	2.405	0.3811	42	1.255

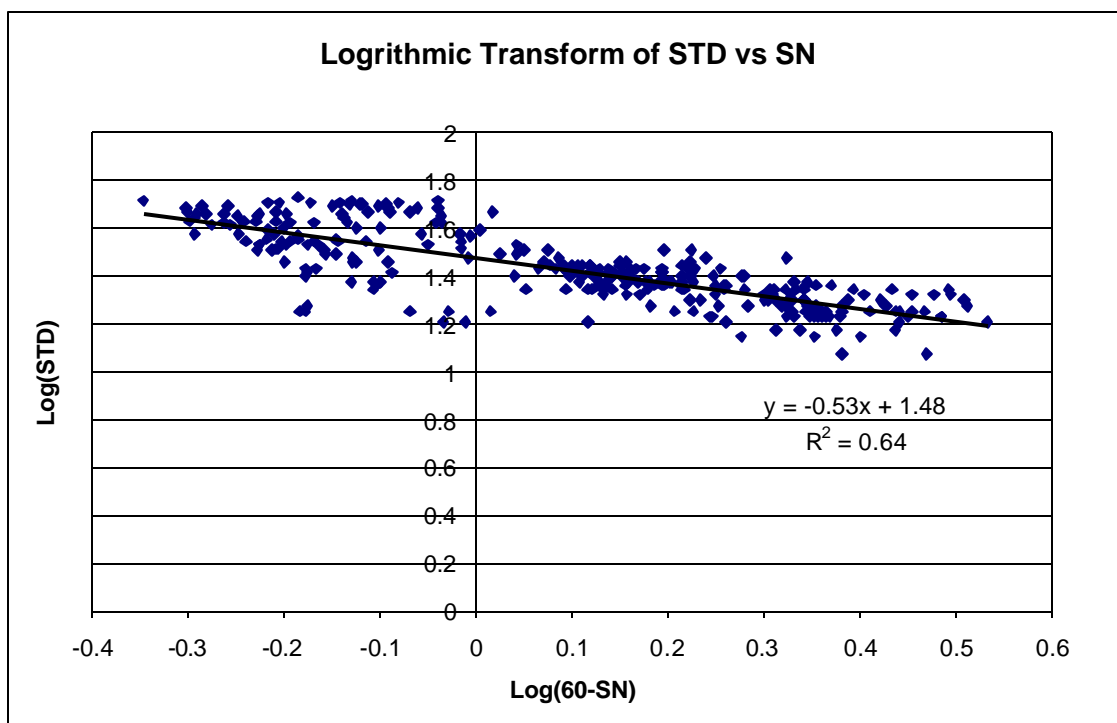


Figure 3.5: Transformed STD vs SN.

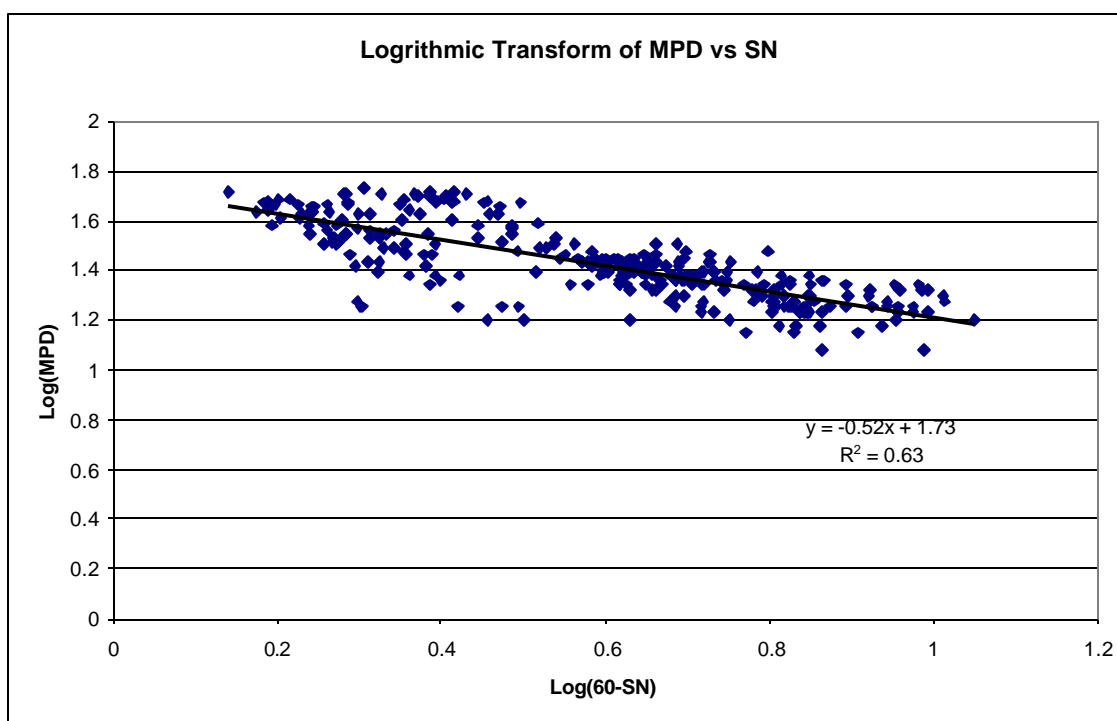


Figure 3.6: Transformed MPD vs SN.

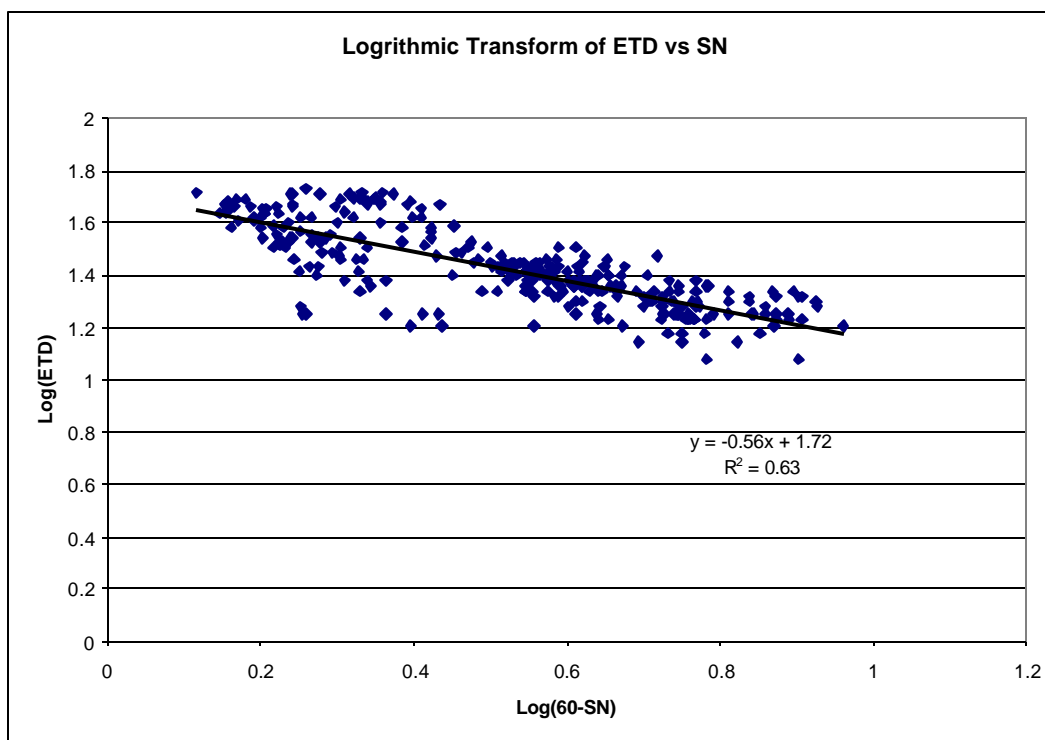


Figure 3.7: Transformed ETD vs SN.

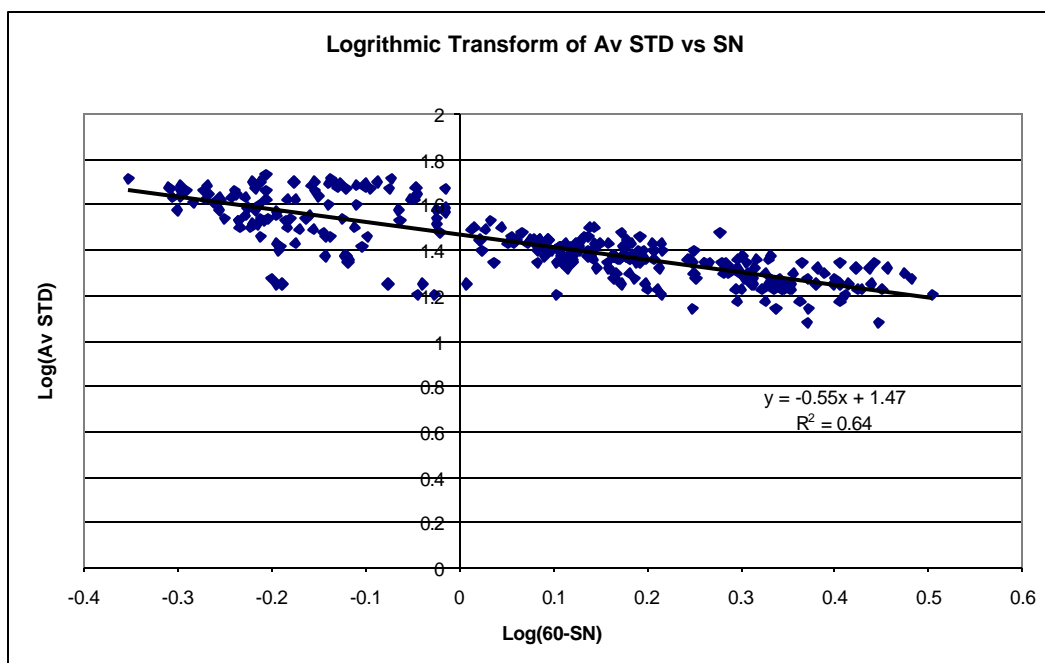


Figure 3.8: Transformed Av STD vs SN.

3.4 Linear Regression Conclusions and Limitations

The best R^2 correlation obtained, as shown in Figure 3.2 was 0.64, which shows that there is some form of correlation between the string parameters discussed in Chapter 2 and the SN. This correlation is slightly better than most other methods used in practice however, it is desired to improve on the correlation if possible. One of the limitations of the linear regression techniques is that one variable is correlated to the SN, when in fact, several variables may be applicable. A combination of these variables will be shown to give better results using a neural network in Chapter 5.

CHAPTER 4

DSP AND POWER SPECTRUM ANALYSIS

4.1 Introduction

Chapter 2 discusses the micro-texture parameters such as MPD and STD. Figure 4.1 shows that the surface contour is similar to a wave, and thus, may have an average wavelength over a section, and may be approximated by an associated Fourier series. Based on this observation, it was conjectured that Digital Signal Processing, DSP, methods may lead to some correlation between the surface texture and the SN. The power spectrum of the data was developed using methods and routines in [11]. This power spectrum was then fed into a neural network previously used in the Transportation Instrumentation Lab of the University of Texas at Arlington. However, the results of this study did not yield good results, and produce poor correlation.

However, it was discovered that the power spectrum of some of the data sets (files) were very unusual in character, with some having numerous frequencies reduced in a pattern resembling a wave form. Further examination showed that some of the data sets may contain numerous bad data points. This could have been caused by a known problem with electrical power during data collection. Thus, the power spectrum plots were useful

in that they could be used to identify bad data sets. These data sets were eliminated in future studies.

4.2 DSP and Power Spectrum Results

This section discusses the examination of the use of the power spectrum of the actual data points describing the surface roughness and texture. The surface of the pavement can be estimated to be a waveform. A waveform can be described by a Fourier series. Each data point in each data set is an estimate of the position of the pavement at that location. This series of points describes the waveform of the texture, as shown in Figure 4.1. A set of such points enables the use of a power spectrum to describe the surface, as discussed in [11].

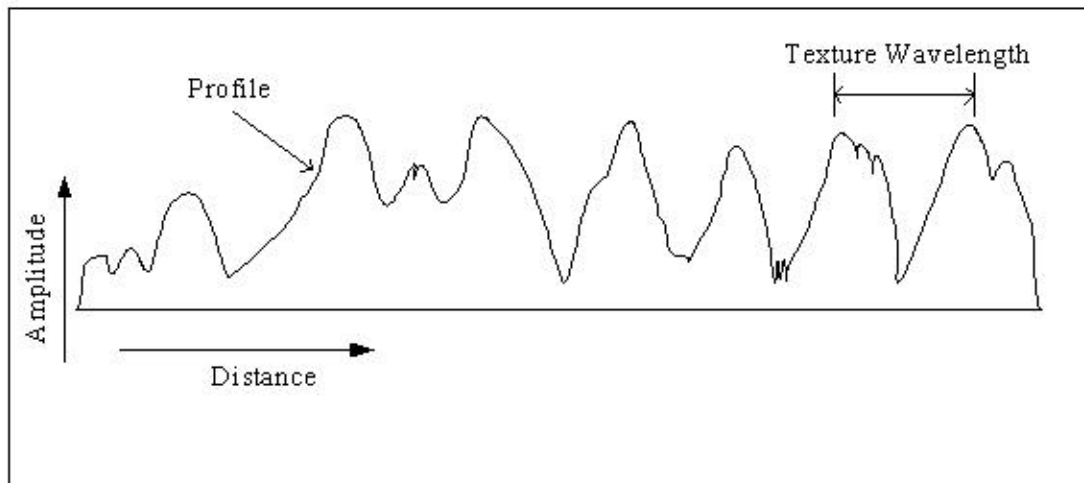


Figure 4.1: Texture described as a waveform.

Using the subroutines and functions developed in [11], a MATLAB program was developed to read the data of each point in a series of data set (set of files for the 02 and

06 data) and an associated power spectrum was generated for each data set. Because of the nature of the frequencies and data, the log of the power and frequency was used. The value of each power spectrum frequency was put into a training file. These values were used to develop a training set and a test set. The training set had not only the power spectrum, but also the measured skid number for the specific training set. Appendix B give an example of a training set. Similarly a test set file was developed, however, the test set did not include the measured skid number.

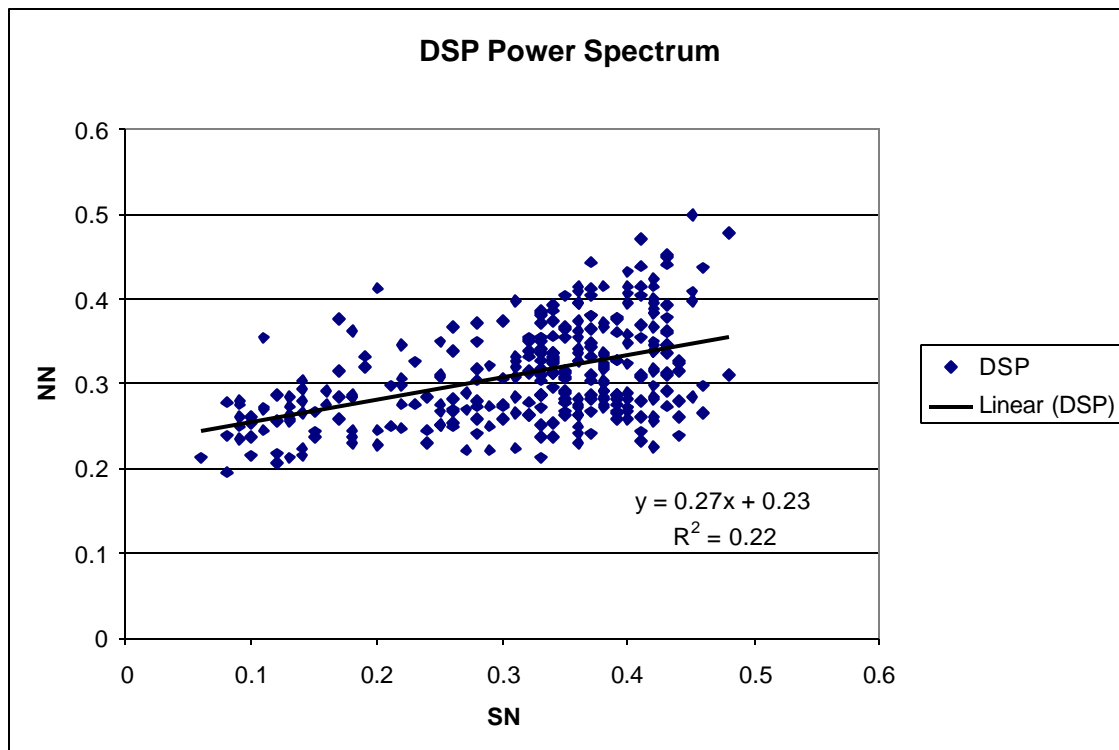


Figure 4.2: Results of attempted correlation using DSP.

A neural-network similar to one previously developed in C, based on [22] was used to test for correlation between the power spectrum of the surface and the skid number.

Tests were conducted on sub-sets of the 02 and 06 data. The results were marginal at best. Figure 4.2 shows the results of one of the tests, giving poor correlation.

4.3 Use of DSP to Reject Faulty Data Sets

Although the DSP methods did not yield good correlation, the power spectrum of individual data sets did reveal important information. Figure 4.3 shows the power spectrum of a typical data set.

About 85% of the files generated power spectrums that appeared as Figure 4.3. This is a "normal" power spectrum and was expected. It was also observed that the string program used to calculate MPD, STD, and other parameters rejected some of the data sets. If the program encountered numerous data points which were flagged as bad, the program could not compute a string section, then the file had to be rejected, and the program printed out the name of the file, rather than calculate MPD, ETD, and STD. For example, if every point in a string section is a substitute for a bad point, then all the points are the same in that string section, and the program terminates, printing out the file name. This was due to the fact that too many of the individual data points inside the program had faulty flags. As shown in Figure 2.8, the data section has 16 bit data. The first four bits are flags used to mark the data and data errors. If one of these bits (flags) are set, then the individual data point is faulty, and rejected. If there are numerous faulty and flagged points, then the entire file is rejected by the program, if computations of MPD, etc. cannot be completed.

These rejected files by the string program had power spectrum that did not look like Figure 4.3. All the files rejected by the string program had unusual power spectrum, but

some of the files with unusual power spectrum were not so corrupted that they would cause the string program to terminate. Because of this observation, the files with unusual power spectrum became suspect, in that it is known that there is bad data within the file, based on the unusual power spectrum, but it cannot be ascertained easily how many bad data points are in the file. The more bad data points are within a file, the more error there will be in calculation of MPD, ETD, and the STDs. However, how big this error is cannot be computed easily.

Thus, the files with unusual power spectrum were suspect, and rejected in further analysis. The 02 data did not have many rejected files (about 10%). However the 06 data had many more rejected files (about 20%) and, in general appears to be "dirtier" than the 02 data. Figures 4.4 through Figure 4.6 show some of the power spectrum anomalies.

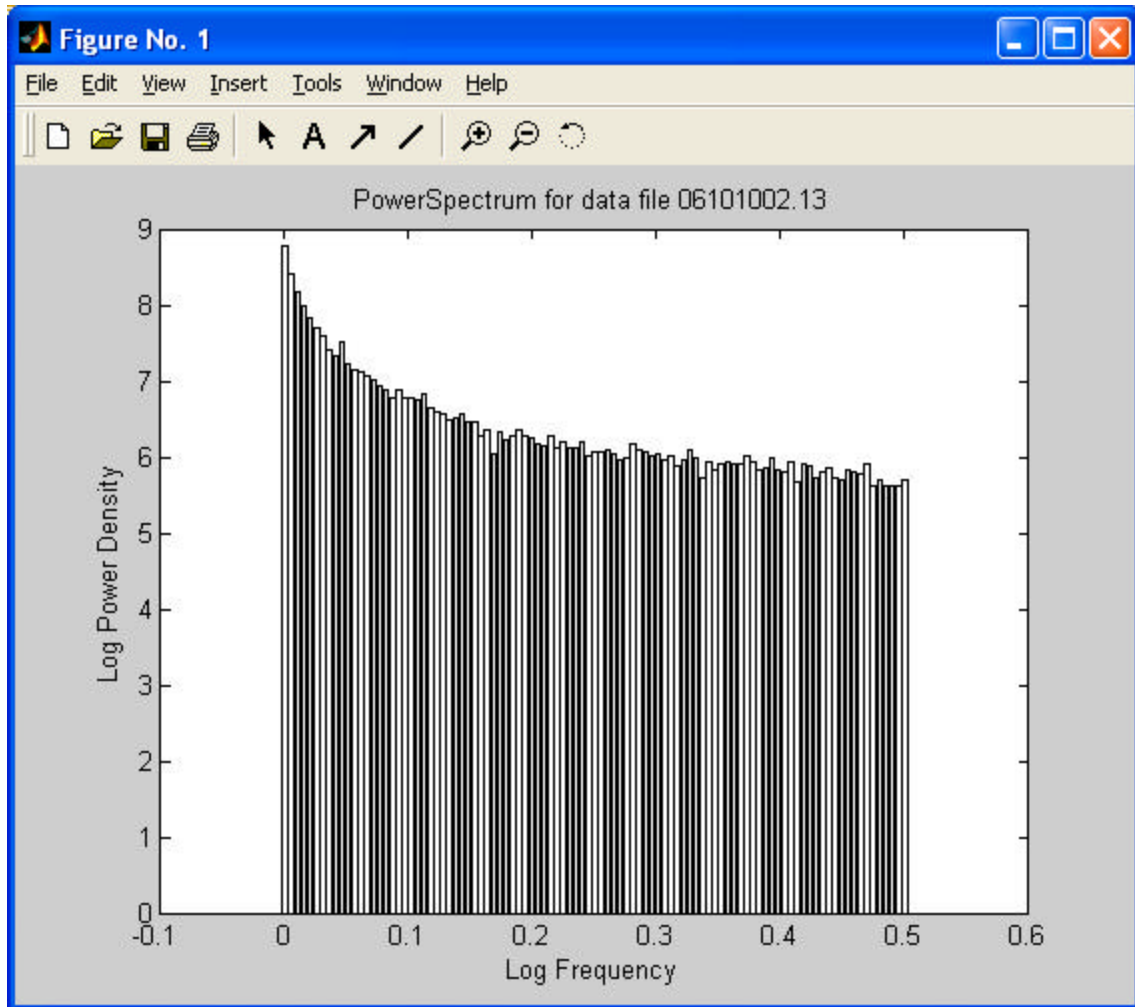


Figure 4.3: Typical power spectrum.

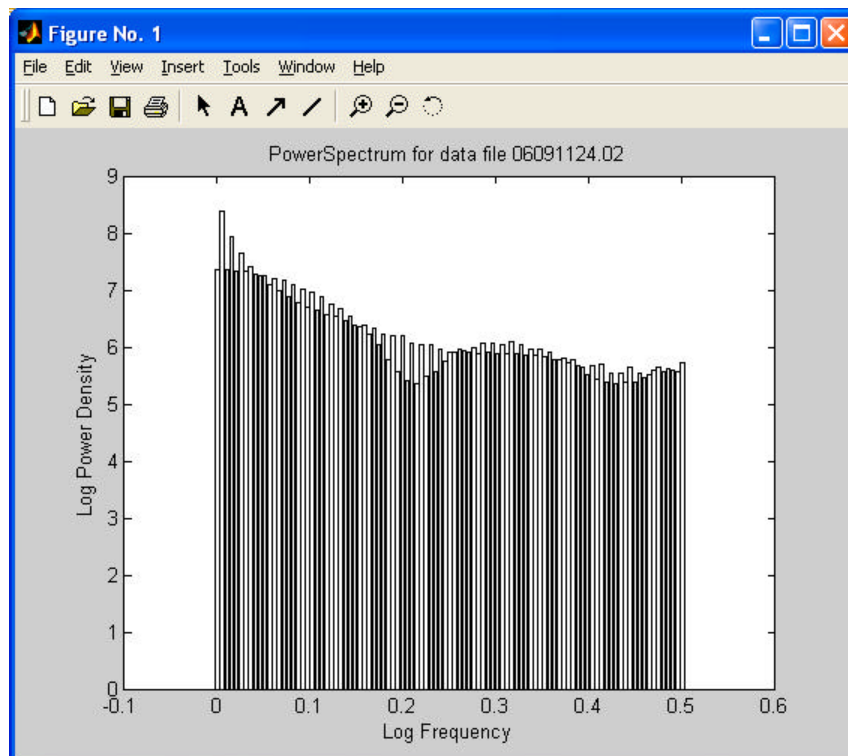
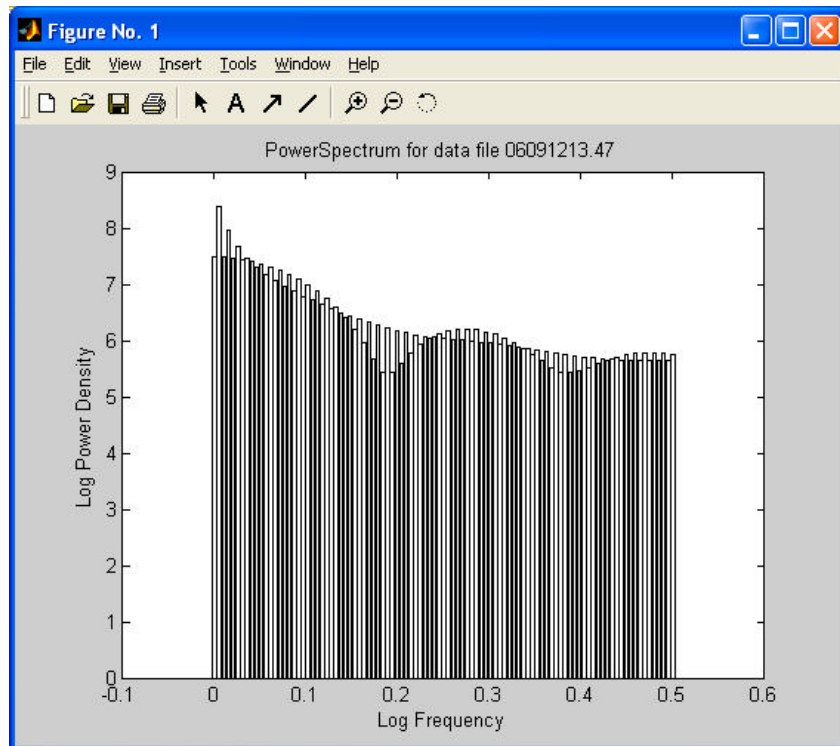


Figure 4.4: Two examples of a wave pattern.

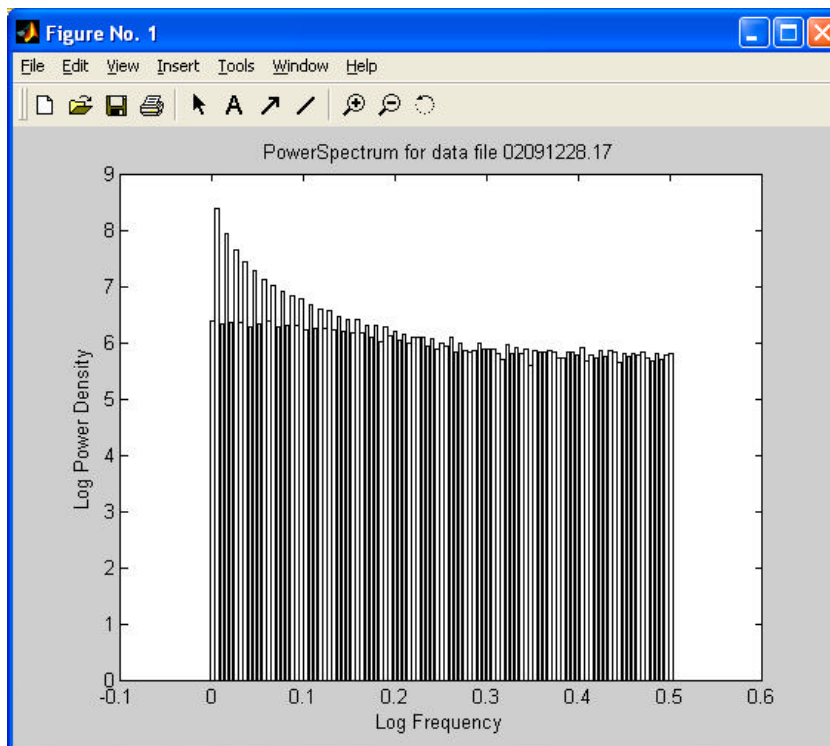
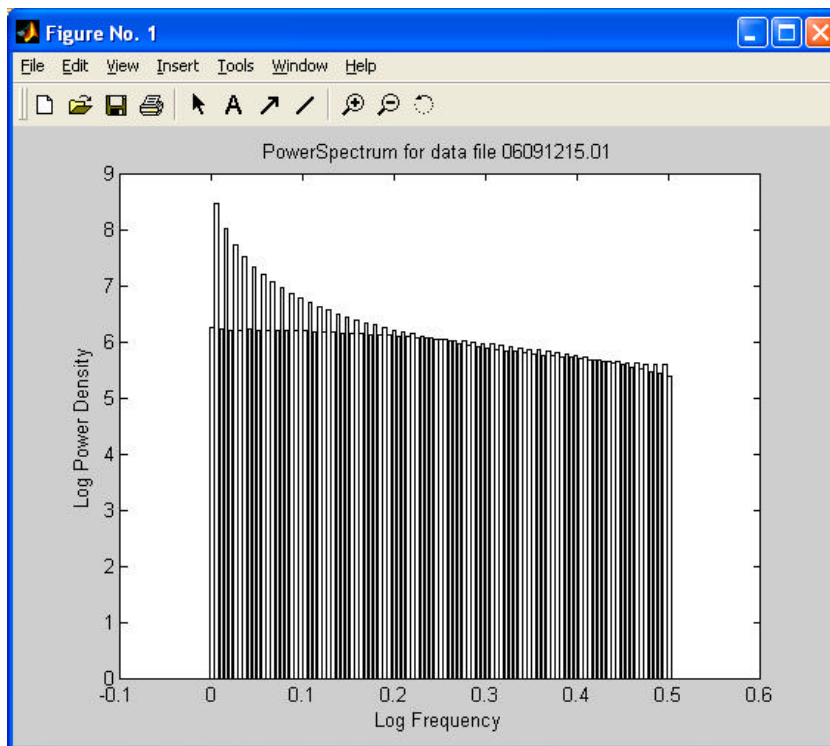


Figure 4.5: A cut frequency patten. Note both 02 and 06 data had these.

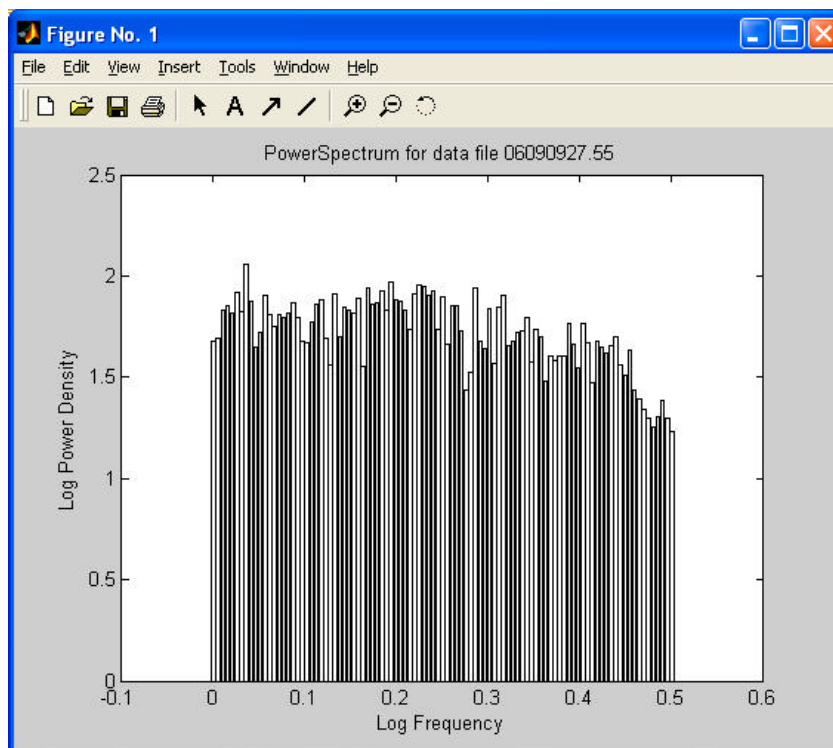
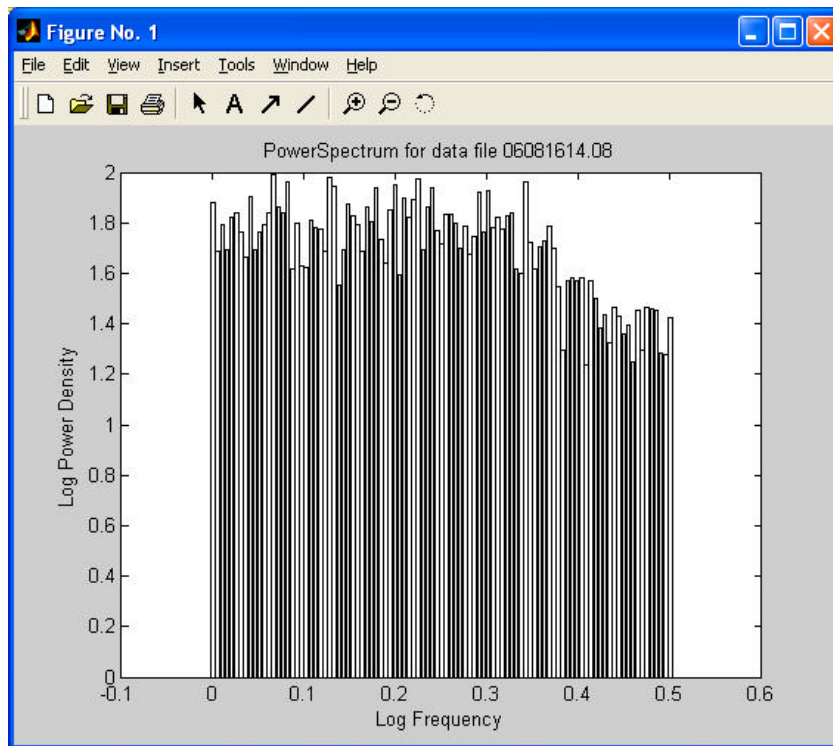


Figure 4.6: Noisy data pattern.

CHAPTER 5

DATA ANALYSIS USING A NEURAL NETWORK AND STRING FILTERING

5.1 Introduction

This chapter discusses the use of a neural network and a so-called string filter to predict and correlate the skid number with the surface texture. A string-line method is commonly used in Europe to estimate ruts and other characteristics. Based on the string-line method, a so-called string filter was developed in [21] to estimate the surface roughness. A neural network was developed that used as input four parameters calculated from the string filter. The neural network was a two-layer, feed-forward, back-propagation network. A training set was developed from all the data, comprised of one hundred points. Two test sets were selected: one from the 02 data and one from the 06 data. Although the 06 data was considerably more noisy and had a considerably large portion with skid numbers above 40, both data sets produced correlation of over 0.70. This chapter discusses the development of the neural network model, discusses the development of the training set and test sets, and presents the results of the neural network, including estimates of the R^2 correlation coefficient.

5.2 The Neural Network

To improve on the process developed in Chapter 3 by incorporating numerous inputs, multi-input neural networks were examined. Rather than use the neural network package used in Chapter 4, the MATLAB Neural Network Toolbox was used. The back-propagation neural-network used in Chapter IV was written in C++ and adapted from [16, 22] and proved cumbersome to use, requiring re-compilation for any modifications. Because MATLAB had been adopted by the Transportation Instrumentation Lab, MATLAB products were preferable. Further, because of its interpretive nature, the MATLAB toolbox enabled quick experimentation with different forms of networks and different inputs. Further, MATLAB enables downloading the network to a microprocessor, so the trained network can be incorporated into an embedded system for future automation of this process.

Through experimentation, it was found that a two layer back-propagation neural network yielded good results. MPD, EPD, Average STD, and Overall STD were computed and fed into the network for each data set. The data was divided into a training set and two tests sets, one for the 02 data and one for the 06 data. The network, after being trained outputs an estimated skid number for each test data set and does not have to be re-trained. These SN were then compared to the actually measured SN. The network used had bias, a log-sigmoid layer and a linear output layer. Figure 5.1 shows a schematic of the neural network used based on the back-propagation network examples in [10, 13, 15, 17], with $R = 4$ inputs, $S1 = 4$ neurons, and $S2 = 1$ neuron.

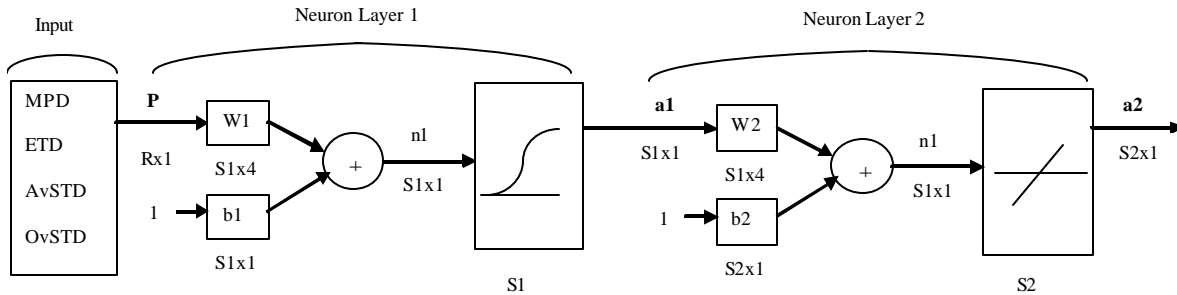


Figure 5.1: Neural Network.

5.3 Neural Network Parameters

During the development of the neural network, several different forms and parameters were examined by experimentation, including a single perceptron, and a linear network. The network of Figure 5.1 using [0 12; 0 10; 0 4; 0 4] for the range of MPD, ETD, STD1, and STD2, 1000 epochs, and `net.trainParam.lr = 0.01` worked well. No significant improvement was noticed using a three layer network.

5.4 Training and Testing Set Development

In order to train the neural network, a training set was developed. In general a training set needs to be arbitrarily selected, so as to represent unbiased data. Initially, the first half of the 02 data was used as a training set, with varying results. It was observed that most of the SNs in that set were between 30 and 40. Thus, there was very few data points below 30 and above 40 to train the network, which lead to a poorly trained network. In general, the training set should be representative of overall data which can be expected to be input [10, 13, 15, 17].

Because of this situation, a pseudo-random set was developed from all the 02 and 06 data combined. The only restrictions on this training set were that it included values from the entire range of SN from below 10 up through 55, and that it not include any data sets rejected by the DSP power spectrum discussed in Chapter IV, so that no faulty readings were input into the training set. The final training set comprised a total of 100 data sets from the 02 and 06 data of approximately 50% each. Appendix C gives the training set inputs.

The training set included two files. One file listed the MPD, ETD, Average STD and Overall STD for the training set, and the second file included the target skid numbers for each of the training data sets. This single training set was fed into the neural network and tested on a set of 02 data and a set of 06 data.

The test sets were developed in a similar fashion. All the remaining data from the 02 set was used, except those that were suspect based on the DSP power spectrum analysis. These data sets and the data sets used in the training set were the only points eliminated from the 02 training set. A few data sets with unrealistically large MPD were also eliminated as discussed below. The 06 training set was developed in a similar manner.

The limits on the input range for the neural network were 11, 10, 4, and 4, for MPD, ETD, Av. STD, and Overall STD respectively. The network performance was greatly affected by these maximum ranges. A few points had unrealistically large MPD and ETD (about 4 times the average). These points had to be rejected because they were clearly out of the range of measurement and appear to have resulted from

an unrealistically rough section of pavement. The total points rejected for large MPD was less than 1% of the data. Appendix C gives sample data.

5.5 Results of Neural Network Skid Prediction

The results of the neural network are presented in Figures 5.2 and 5.3. The 02 test data resulted in a R^2 correlation coefficient greater than 0.75, which is very acceptable. The 06 test data resulted in a R^2 correlation coefficient of 0.71. This is a significant improvement over the commonly used sand-patch method or the linear regression methods developed in Chapter 3.

For the 02 data, over 200 data points were used to test the neural network. These points include all the points that were not used in the training set and were not eliminated by the DSP methods discussed in Chapter 4. Thus, the test sets were approximately twice the size of the overall training set. However, it appears that the training set did cover a good spectrum of possible data values.

However, there are several observations that may adversely affect the results. First, the neural network tends to slightly under estimate SN high values. The reason for this is unknown. Second, the network tends to over predict low SN. This may be caused by faulty measurements and low velocity of the vehicle and needs further investigation. Further, it should be noted that around measured SN of 50, in the 06 data, there appears to be a flat line in the actual test data which is visible in Figure 5.3. This line may be a limit in the reliability of the measurements above 50. In general, skid numbers below 30 and above 50 are suspect in that this leads to a very high or low friction coefficient.

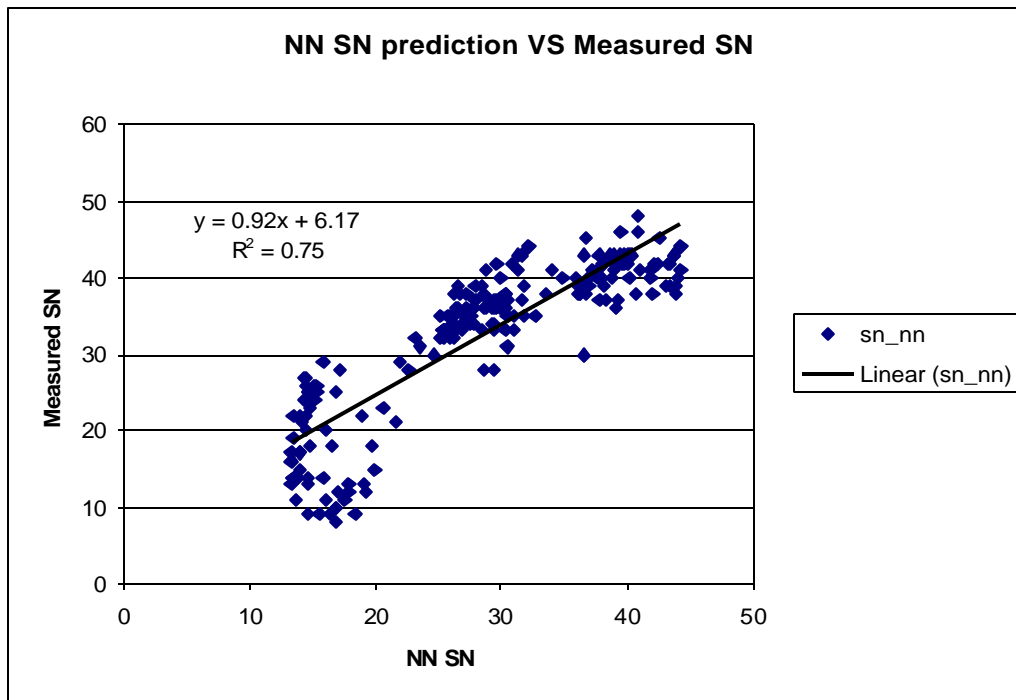


Figure 5.2: 02 test data results.

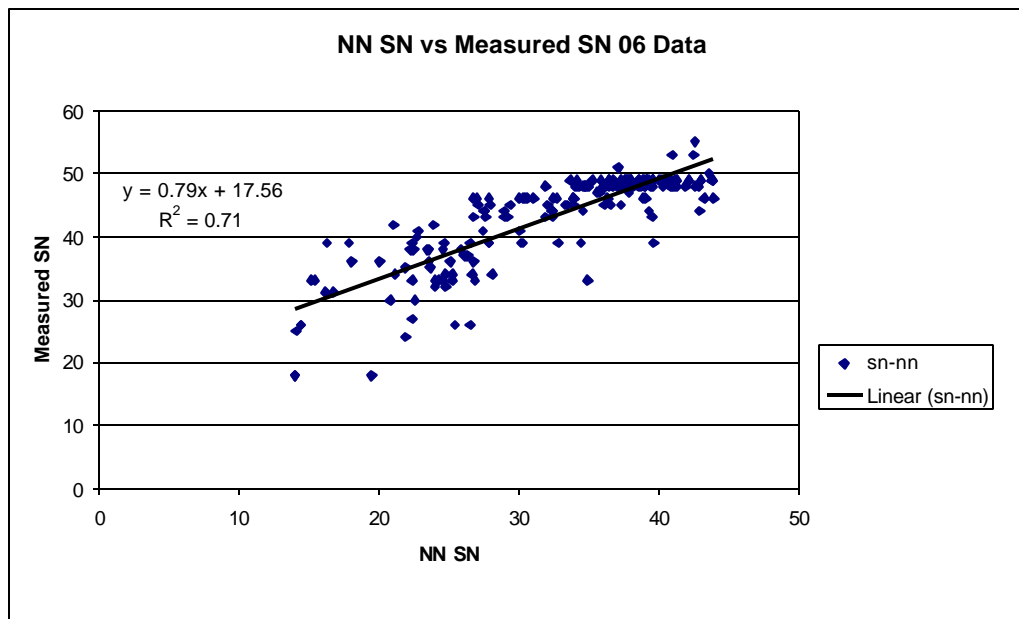


Figure 5.3: 06 test data results.

5.6 Further Neural Network Tests

In general a two layer, back-propagation neural network with log-sigmoid and linear functions can be used to simulate any function [17]. However, because of the highly non-linear nature of friction, a three layer network may produce better results. To test this hypothesis, a three layer back-propagation neural network was developed using pure-linear, log-sigmoid, and tangential-sigmoid layers. Results of this network were only marginally better than the two layer network presented above, and results are questionable, and will not be presented here. Further investigation of this and other matters will be discussed in Chapter 6.

CHAPTER 6

CONCLUSIONS AND FURTHER RESEARCH

6.1 Conclusions and Summary

This document presented results of data analysis for the prediction of the skid number between a rubber tire and pavement. The skid number is related to friction. Chapter 2 presented background information, including an overview of the hardware used to collect the data. Chapter 3 presented a linear regression model used in the preliminary analysis of the data which gave acceptable results. Chapter 4 presented DPS and power spectrum analysis of the data, and although this analysis did not result in good correlation, it did provide a method of checking for faulty data sets. These data sets were subsequently removed. Finally, Chapter 5 presented a neural network which was able to incorporate a number of inputs from the string filter of the pavement texture measurement. This network was able to predict the skid number and give a correlation of over 0.70, which is an improvement over conventional methods. Table 6.1 summarizes the results of the Linear Regression methods, DSP, and Neural-Network methods.

Table 6.1: Results Summary.

Method	Data Set	Best R^2	Comments
Logarithmic Best Fit	02	0.60	Used Overall STD vs SN
Linear Regression	02	0.64	Used Logarithmic Transform
DSP Power Spectrum	02	0.22	Used C++ NN routine
NN-Matlab	02	0.75	Used Matlab NN toolbox
NN-Matlab	06	0.71	Data was noisier than 02 data

6.2 Issues, Concerns, and Areas for Further Research

Although the neural network developed in Chapter 5 did produce good correlation of the data, and did incorporate a number of inputs, two additional pieces of information will probably lead to better results. First, it has been shown that the measured skid number varies with velocity [3, 4]. Although the ASTM specification defines the velocity range at which measurements are to be taken, there will be some variation which will affect the SN measurements [3]. The 06 data includes velocity measurements, however, the 02 data did not have velocity recorded available at the time of this research. Because of the lack of velocity data in the 02 data sets, velocity was not incorporated into the model. The 06 data shows a substantial shift in velocity in some of the measurements. This probably adversely affected the results. Because of this shift in velocity, more data should be collected, recording the velocity of each data set, and incorporating this data into the model.

Second, the 02 data was collected in February 2000, and the 06 data was collected in June of 2000. In all probability the temperature of the 06 data was significantly higher than that of the 02 data. In general, rubber durometer changes

with temperature, and the durometer of the tire rubber, and other rubber characteristics, may change slightly because of these temperature changes. This change in durometer probably increases the friction coefficient. In neither data set was the temperature recorded, however, the 06 data had higher SN values than the 02 data. Thus, the tests should be repeated, recording the temperature, and incorporating this into the model.

Thus, it is recommended that further tests be conducted including incorporation of temperature and velocity.

Further, each data set should be examined by the power spectrum analysis, to check that the data set is good, and does not have an unusual pattern as discussed in Chapter IV. Data sets with an unusual pattern can be rejected on the spot, and the data re-taken. Then immediately the data sets should be entered into a test set to see the effect of the specific data set. To run the power spectrum and neural network takes only a few minutes, so this data should be collectable and analyzed in near real-time, and the prototype system improved.

Eventually, if the system is proven to reliably predict the skid number, the power spectrum analysis and neural network could be incorporated into a near real-time system to quickly estimate the skid number of a section of pavement. This research has laid the foundation for such a system.

APPENDIX A
HARDWARE AND SOFTWARE SCHEMATICS

Illustration of the Measuring & Data Processing Procedure

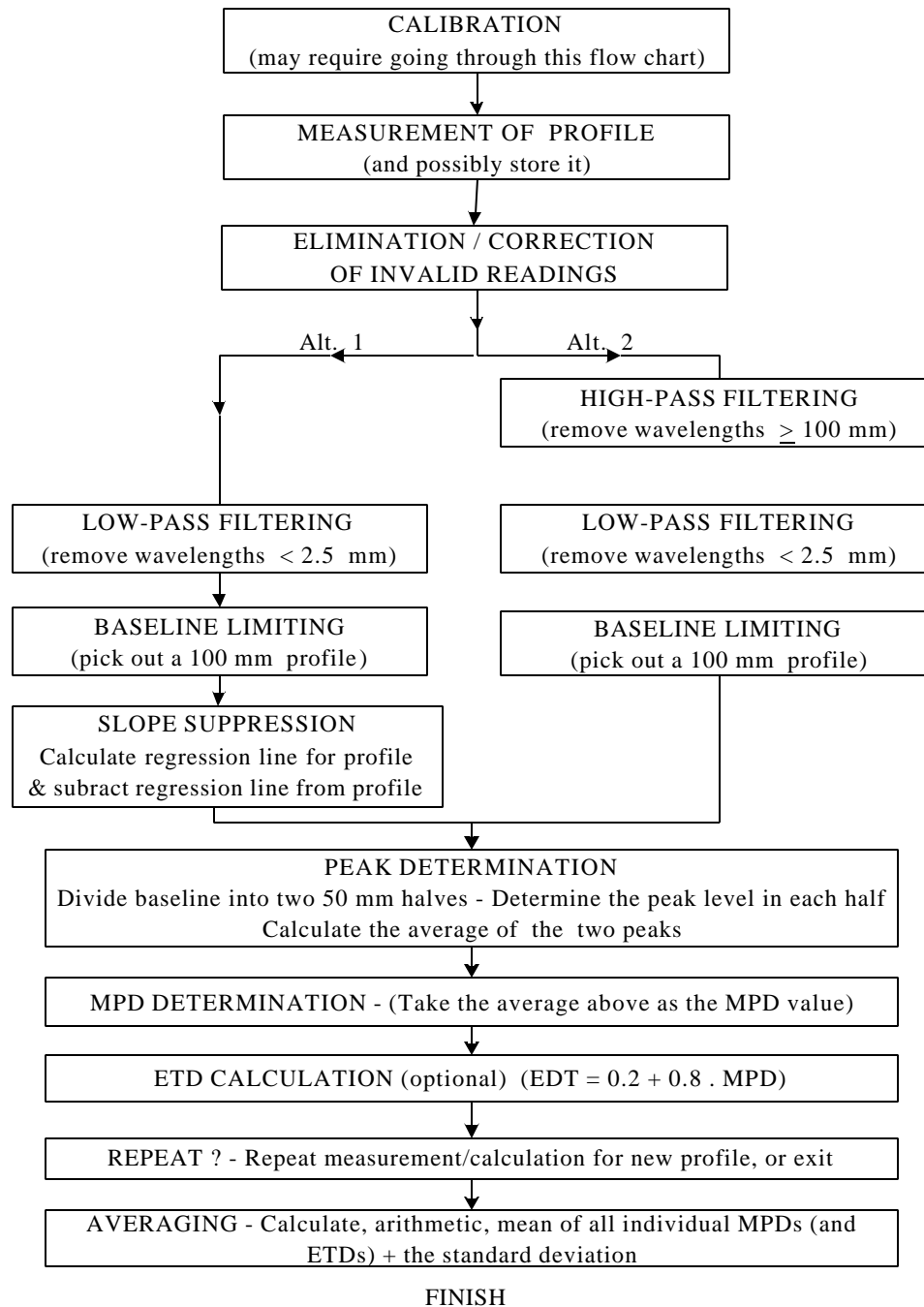


Figure A.1: Measuring & Data Processing Procedure.

APPENDIX B

DSP TRAINING SET

Below is a sample training set using 62 power spectral frequencies. Test sets were similar except ignored the input skid number.

SNNS pattern definition file V3.2
generated at 20-Feb-2005 16:7

No. of patterns : 100
No. of input units : 62
No. of output units : 1

```
# Input patern 1:
66920187.897793 32941035.569412 23437597.363091 17102669.264975 12375282.247373
8706451.995003 6902257.775912 5908976.134060 5890983.809944 5646777.717170 4333716.456347
2292135.818704 2893704.064684 1774189.646392 1828983.203342 2807090.107007 1940404.912016
1380705.932342 1476759.050407 1368184.258929 1957798.622319 1087046.631661 1118246.402403
1289113.032590 1394383.808489 1245390.786830 1191570.296332 765733.030430 1030570.075107
906982.449463 899895.012009 735405.675547 774827.877264 781802.303863 474934.413461
1143384.548680 574815.551448 622722.848649 501121.132508 651019.557172 988438.234749
637251.202498 719121.992870 610071.409229 507990.605722 837105.327730 637186.687494
589613.044104 420922.006586 648077.185956 697637.865367 569259.932254 606596.329565
401638.377226 505787.123896 531308.547966 620212.512861 872843.745290 382565.283167
593640.297812 601466.106947 642702.435303
# Output patern 1:
0.530000
# Input patern 2:
72116228.322030 35256411.809467 23889382.793937 16442250.898121 12320555.427253
10408961.377553 7578640.876279 5572973.341398 5221509.381249 4548786.173249
3508466.764558 3090598.816205 2779208.940324 2493975.317202 2398684.705368 1700545.199604
573263.043070 603279.431756 1467085.560972 1501622.055320 980398.201004 959636.858664
1143522.039794 1063010.571967 1007726.029697 992352.903516 836693.010459 795727.465782
887503.341376 825020.502930 659459.092813 579530.717905 534115.534929 546794.903084
613020.192379 521118.059434 393424.289735 499288.687116 574094.592098 388323.955150
303738.986894 466817.258709 513273.959108 415969.933558 419150.592183 454869.938658
435298.477053 480587.674676 533313.620293 443575.062534 316125.523720 260245.427567
263350.950964 353625.041511 461828.704032 439928.217365 391535.618526 409787.899091
352972.754191 256129.069596 331115.567509 427806.744629
# Output patern 2:
0.500000
# Input patern 3:
60475628.032078 38630686.895999 22476452.448680 15581979.882860 11713849.632954
9412418.067209 6816016.470384 6659688.509732 4859393.415192 3449405.549713 4084114.920168
4673561.140919 1120946.134910 2576155.031441 1501074.446773 1416385.521160 1879945.435833
1362767.247748 1593565.049011 1508073.712519 1004977.300976 1152436.387032 1181396.614479
1426370.642061 392768.876775 1209332.600148 1044163.643738 916474.531049 909891.232913
750550.748779 603522.618433 690351.172314 403666.872300 626481.924636 823204.010805
525759.553767 418771.523209 602276.372721 754702.134233 352691.311116 584161.302744
486582.158788 589602.988099 534686.666258 559383.663111 521971.883110 694752.848575
247123.735443 338695.007599 557420.060625 470553.531428 337301.798729 389925.006875
531757.443546 371691.749142 440165.807813 365153.034302 439123.322110 511932.152318
252765.329124 519438.652659 625965.795654
# Output patern 3:
0.490000
# Input patern 4:
70174582.993319 39093821.283944 23669892.457786 15721680.711944 13632358.638952
9614617.278474 6821853.274720 6159218.964947 5098552.985648 4301955.543288 3262601.747185
2655236.487536 3131281.503928 2144645.513828 762119.532989 1313636.947285 1836758.105747
1657079.600178 1790310.473925 1491765.086250 1061679.369729 1221444.150466 1172459.657556
950911.442282 917336.539136 875613.669858 876439.748800 791528.312755 675555.040616
```

```

783217.566406 717513.393156 576710.495809 754478.397933 689416.629007 430476.254925
513727.629084 557292.276118 470773.898003 547920.070089 526203.209793 451010.162337
518570.265225 508726.284731 464989.000964 468217.782982 476660.314297 515669.694483
447936.413783 405441.411382 580065.757302 544337.642222 265718.557141 207441.191743
254070.346885 328206.411508 495852.592845 461769.468945 341453.512917 426005.964743
476705.025023 416157.151656 394135.232422
# Output patern 4:
0.510000
# Input patern 5:
72286651.190605 35946173.957656 24324405.422707 18777694.801473 12281638.191813
9697670.754125 8822141.247413 6910371.560205 6185598.159410 4615314.488581 3987046.909205
3853311.683962 4054572.873900 3894614.313370 3570143.725923 3122631.373083 991955.814019
2161350.871417 1696322.093905 1879310.918911 1282944.597204 1063528.713185 1216759.788428
1270512.882696 1112921.555665 1332420.495343 953939.659496 894706.381836 945888.957852
873188.584229 893016.897892 1101883.888025 833508.018569 798674.122297 750951.824066
908611.885612 665516.395080 861441.454424 551588.678899 588099.447972 813404.614349
1037321.511149 720100.995449 896345.488004 884174.729567 798821.369735 640254.251905
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645164.905414 689557.634803 395012.186588 717172.920725 597372.878852 338588.002651
478558.630814 475458.980748 312909.433838
# Output patern 5:
0.490000
# Input patern 6:
73575330.607193 36800742.236098 29179667.599438 17983378.379822 12111765.800753
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933211.764819 859721.872300 738362.388026 1061473.711669 600012.991959 710669.848578
711643.384725 566986.791553 586186.097656
# Output patern 6:
0.360000
# Input patern 7:
59058897.446602 34211278.634336 23447771.349545 15363384.877910 12733639.680787
8958988.565974 7593632.998850 5835269.728988 5194587.327092 4119154.667313 4405012.842305
2917080.185443 3261574.376713 2664459.472813 1935757.304539 1815711.404272 1967808.435366
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495044.874148 546922.477038 285972.919818 455739.756096 391513.502226 407094.593944
417586.501394 274315.527159 280329.433400 348552.925833 273819.621385 328454.084820
273412.956356 423082.212910 345487.431152
# Output patern 7:
0.500000
# Input patern 8:
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9741485.119405 7223022.834401 6230177.139779 5572004.944787 4270632.548577 3659084.817829
3251123.454253 2560342.360649 2573593.675677 2796858.714876 2305314.029161 2039975.500124
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289665.745725 306279.619905 377921.371094
# Output patern 8:
0.500000
# Input patern 9:
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332606.011234 263436.060111 197075.171631
# Output patern 9:
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# Input patern 10:
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343241.133740 360115.603603 375207.796277 360415.382936 349127.812559 372719.833808
399771.568077 379473.300733 313933.014696 280076.068604
# Output patern 10:
0.500000
# Input patern 11:
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509285.812687 390007.667019 560572.561318 411210.116166 547092.795920 332099.538439
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# Output patern 11:
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# Input patern 12:
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# Output patern 12:
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# Input patern 13:
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213755.452203 345479.347513 406347.379512 327004.779521 355840.237128 299230.179983
263573.142845 351652.469414 393740.571777
# Output patern 13:
0.490000
# Input patern 14:
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# Output patern 14:
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# Input patern 15:
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# Output patern 15:
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# Input patern 16:
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# Output patern 16:
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# Input patern 17:
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# Output patern 17:
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# Input patern 18:
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432276.124048 331020.939635 443816.839111
# Output patern 18:
0.490000
# Input patern 19:
92056994.610428 36675331.723424 24850797.830839 20970322.616687 13420280.868459
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# Output patern 19:
0.480000
# Input patern 20:
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385344.316663 341813.773288 383857.340032 246799.667120 409810.574833 362614.108225
219579.033246 396686.138351 387293.510881 195932.075439
# Output patern 20:
0.500000
# Input patern 21:
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283393.141285 254196.295531 235990.649181 230689.235352
# Output patern 21:
0.500000
# Input patern 22:
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441653.400747 360005.715210 340872.730595 395957.989338 380577.952108 385593.399604
399072.417130 377650.378835 346576.684082
# Output patern 22:
0.480000
# Input patern 23:
5763927.535103 57169063.436888 26194789.745976 4787425.252732 20302446.988056
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6392315.877278 2514222.553764 2938009.735816 3941089.054390 1497643.686408 2114346.929743
2520514.541327 1293673.680201 2137383.614499 2048192.594743 936170.916896 1598913.340543
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597518.943389 369047.939300 346654.834087 517232.682373
# Output patern 23:
0.490000
# Input patern 24:

```

```

71342801.752792 19523794.145752 24513040.197154 19486242.389785 12454107.783913
11685495.944854 9695345.073458 5626889.857572 4853314.210900 4985462.082717
4401660.750400 3771694.840034 3601607.249590 3098728.028704 2661166.121150 2337638.728795
1961618.432248 2096384.031404 1910476.864238 1808224.189630 1616234.584346 1658061.025596
1342799.795612 1374025.527361 1517326.975896 1296394.892296 1066797.458127 975635.722378
983316.728104 799316.872070 920008.571165 911692.760031 844878.953159 791084.282668
837501.664437 769664.046390 626741.933817 561038.258409 575071.886116 667021.936110
699330.888876 647468.311603 649359.125648 616360.001223 552632.238104 599335.336530
666591.244221 554486.466488 423020.051817 546172.973425 527041.891661 512233.583309
528894.067137 609639.294579 495555.524870 505544.754914 468081.921705 338795.506697
514683.210975 688810.871301 528265.941216 212337.154297
# Output patern 24:
0.500000
# Input patern 25:
107844826.510076 73382253.210985 45548281.853734 22625563.930538 8140118.367469
2159600.318738 1705396.073801 3876815.937193 6637253.732170 7145898.171160 7156375.399136
4633853.065122 3616072.497778 1660730.339637 1306786.875142 1474216.187212 1836178.409474
2442992.189783 2737408.465756 2027263.259247 1913277.446492 1198440.671179 1266608.823297
1043494.601086 1119396.162022 1205776.400693 1300602.798194 1543391.965444 1284067.872849
1080366.774658 1080290.178747 866406.152413 858905.704911 713303.525554 557870.285602
755342.208228 735484.792698 876567.635769 802442.712828 841721.552659 746772.295542
671114.596071 509603.904580 549621.224525 502129.899260 638281.612023 729852.322429
817417.061359 759822.290511 694777.977121 479089.051541 489279.802004 469655.633628
501801.482713 528057.858832 699145.526645 827399.834681 741875.629805 628903.574977
405477.659719 313308.253848 428100.053955
# Output patern 25:
0.500000
# Input patern 26:
69434366.695234 40061937.428863 23337480.870888 14283782.248930 11576759.105180
7914096.869289 8321910.656852 7380239.202397 4870356.674679 4410822.930195 3588796.721143
2961118.638923 2689127.507245 2184074.562289 2352151.779714 2227936.848859 1802927.513454
1722615.632797 1367600.334148 1184568.619887 1270365.334717 1195016.368907 1153342.507965
1075483.347491 947888.213895 840097.593803 877319.018364 851619.924983 518652.258246
721203.356201 1042125.873162 553034.598767 445279.739928 700976.020876 525164.705245
642983.631014 667267.313969 320390.810403 408514.660979 579810.862150 596829.259089
547040.843384 317906.503162 381366.573840 464789.331427 317637.126676 470980.806518
584922.264057 421034.209793 289562.561997 236804.254002 426008.394160 506980.733022
302496.456284 379245.977122 478679.414526 318014.336956 305018.345508 347656.603437
315230.319173 387688.919768 460478.989014
# Output patern 26:
0.490000
# Input patern 27:
4902640.269267 47094325.609103 38646677.113494 4535967.211999 13677994.030292
14661257.157788 3763009.973586 6509091.503508 7007651.989817 2856571.207996
3947197.752572 3668339.425147 2147763.863870 2687787.905177 2318198.392860 1727957.009679
1979531.280315 1878311.456938 1390373.957330 1464301.027675 1573988.667670 1070122.307866
1029287.406623 1211332.390326 935632.147894 726930.299715 955199.949003 995755.146154
565986.133521 835632.120850 1041060.582124 466965.679300 731846.106170 887246.651694
413372.179121 583458.603728 697106.461529 430533.489496 495004.610340 624994.908445
479669.546675 501294.407877 625307.459500 443369.283033 526306.993042 557772.376562
333814.689271 522322.328589 443578.766299 287805.063346 513319.254787 423678.417537
343171.933643 503653.907154 482159.016744 370322.249498 466643.947273 503809.910752
299903.159946 432046.523107 456919.039910 246320.418213
# Output patern 27:
0.510000
# Input patern 28:
1590971.808800 1641049.347590 1658285.337371 1617497.071217 1589090.602228 1572663.653667
1571109.171021 1559639.724581 1537289.783642 1503056.021566 1482543.952524 1457542.761012
1434756.994073 1413086.459327 1380438.641658 1349165.130453 1320571.338549 1287726.322996
1252857.320581 1223058.028380 1188180.370646 1153267.269076 1120288.350454 1081922.551203
1044808.481257 1004383.708665 967674.164903 927031.493817 893735.267481 856504.495850
818090.731291 778514.975466 735331.233929 701227.620585 660723.988970 625264.683522
587865.723952 552073.418935 518881.944983 483403.468384 447157.168985 416156.327545
384375.810414 352507.999213 322923.033165 297888.437646 268453.920237 243817.583481
220074.760534 197832.751726 177433.746436 157976.577116 140673.798972 125585.055783
111545.141758 99607.073426 89560.505292 80937.565342 74333.555472 69313.807710
66657.487758 65865.435791
# Output patern 28:
0.500000

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# Input patern 29:
95466344.895484 46664786.499546 24482219.460659 18162860.500075 15138047.265266
10769862.427187 6953003.679816 5616377.857418 5752700.945355 5266459.004586
3891406.304759 2780159.753415 2537082.896677 2622132.609990 2299975.707936 1722108.473182
1430853.458685 1475980.335689 1490306.658076 1279341.711781 1063070.171350 1085753.566863
1212839.926387 1185736.444580 1006084.535435 861150.888457 864651.144061 885910.432730
828590.837821 716468.261719 662515.518648 641951.925300 594919.689423 550975.409781
576608.490531 651690.484728 644687.591095 519846.598396 409653.231614 444719.493027
559620.032814 563252.241837 409104.306162 252989.801387 260214.196535 397066.011104
483244.987177 439134.830554 356027.794989 353850.493290 438037.023720 516930.058698
531040.731403 487995.641929 428584.834732 357733.822264 281394.587975 226913.392439
239662.834095 320507.796239 417505.474226 456929.131836
# Output patern 29:
0.500000
# Input patern 30:
40640399.081410 26739518.229281 29079694.578517 16432963.169742 13388956.119184
9537542.611696 6118981.644137 6691266.842871 5479593.445494 3804487.698690 3838604.300087
3370133.427849 2488010.950330 2299236.545567 2486224.083648 2092604.203688 1501901.299276
1610474.421367 1623404.533854 1280219.176071 1291308.399779 1233642.065029 1004576.454790
975513.341070 967245.718259 999662.411710 873211.644786 615799.595972 790335.192856
846743.187744 614546.410333 696276.674128 595044.016002 455390.045423 779441.713861
633946.536233 296223.763128 586353.945134 619790.147762 366613.883413 536993.173190
529436.888928 312956.131778 434516.910883 546563.338901 448243.979336 316813.333098
324786.526468 508087.069612 487088.112903 281103.445755 311766.282012 438663.648147
439968.750211 350191.516722 290588.157718 393813.324057 422427.025682 313957.494314
360030.458827 386040.169943 328397.405518
# Output patern 30:
0.490000
# Input patern 31:
90095811.143481 36357333.702579 25133590.913130 16599317.498158 10646376.898309
8948982.107070 6737410.238878 5162603.323313 5703410.920240 5341408.419297 3701531.760233
3071954.036118 2961716.012708 2239948.769697 1884291.881950 2073384.067935 1818737.521786
1472804.652269 1603106.459012 1554487.940519 1189817.318625 1120548.940128 1178892.924368
983239.329718 839307.601592 946646.525070 918580.521551 716244.913806 665045.494532
790999.553955 842205.534058 700721.522588 548167.757759 551966.764090 566363.981322
484912.372032 518299.002438 679132.108396 610066.134515 363664.011279 364566.865391
540528.482472 525208.648250 432662.526106 431885.751637 383050.025713 311522.636112
377163.353898 483303.169574 510889.700320 485548.187238 338476.556614 182606.472817
301068.409835 496425.627395 390824.676004 265362.896653 430530.552731 537386.568037
372731.546576 255686.166412 261379.268311
# Output patern 31:
0.500000
# Input patern 32:
108753795.846869 22275788.960872 29535856.578366 14883797.017175 12906483.997533
8977843.632233 8281030.629394 5186121.485259 5916612.138993 3606839.040989 4064141.443672
3093135.951706 2638882.600573 2761037.100300 1898410.481519 2182012.285737 1702665.543460
1566169.175817 1650519.263887 1136486.032992 1494027.301308 962614.550149 1237111.112001
897666.585440 1063581.103868 773714.466863 1007360.959086 590978.685351 1012307.248166
458710.909668 939570.866145 444749.031182 788280.006200 502474.986841 628560.825840
542774.993407 542474.943089 517364.696424 522498.619690 472538.987623 495093.323042
457793.241689 430890.325530 491317.099556 347860.091242 528096.640911 300955.450331
510715.004057 322347.365576 437470.364532 382227.061881 359717.238004 422514.132486
328564.046976 408847.347343 351455.408146 358304.660904 401404.879585 301058.858913
442520.381464 266476.261543 459237.710449
# Output patern 32:
0.500000
# Input patern 33:
58807558.503916 40185396.596139 19478962.613905 15908269.641382 13180487.428941
6056208.519614 9733990.576554 7614214.082082 5165319.452595 4024718.031825 4723443.322005
4015723.223745 3027737.503536 2970685.767263 2722682.045770 2363282.791601 3029954.034506
2127530.277299 2246226.629019 1755327.135460 2130667.185285 2019062.677454 1402315.513649
1432286.429929 1386149.673625 1480720.053840 808148.689730 1162857.062932 1222441.755818
1162176.864746 894125.412786 1037936.260796 882756.902001 1132080.895068 707765.289823
894668.948179 1120219.569111 930800.691272 854169.924178 970193.922082 743323.854267
749932.476021 555454.888123 814062.423541 698766.734933 757465.917307 882504.342948
757812.995620 735939.373895 714391.647468 658482.889996 549632.246528 681095.172091
516629.115567 606566.689790 564316.049311 523340.644113 729253.049683 387468.806283
705026.467713 732075.713476 536389.610352
# Output patern 33:

```

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0.460000
# Input patern 34:
68949144.284723 29285320.278595 19206360.575513 12499130.480441 11798509.863182
9445401.213103 7050943.094257 6118079.051012 4358389.871411 3530138.433074 3446968.950845
2843484.417610 2597151.410827 2349377.060331 1902464.812637 1909632.724444 1754032.100852
1419293.544801 1248642.849495 1114927.878086 1282240.437705 1305610.731200 896560.296072
888137.908148 943804.456489 698824.170910 811733.225446 843286.598699 598676.674266
804602.790771 777183.814203 336390.497429 504519.650009 757972.622126 531879.025569
523725.706378 605920.349966 417616.453179 348717.297826 469277.826835 558298.315611
492880.231674 350945.464603 414655.823701 481999.715539 307984.198946 276376.566333
482097.888125 544106.019607 381859.135580 202511.398737 265466.214425 462047.627390
443101.290995 360360.334160 341186.890719 252393.872673 335351.495349 461821.446888
313347.914437 293050.149364 411064.770264
# Output patern 34:
0.500000
# Input patern 35:
53187876.415728 25270281.057211 18461795.969128 11066157.050314 8778908.108721
6444432.137086 5197059.613983 4611638.893283 3524511.367546 3603268.026991 1998254.163978
4001180.305847 2714335.984686 1628806.027537 1508728.280297 1441971.382928 1276230.539884
1372666.027042 1181400.158107 1192795.762427 950495.966425 900790.621759 729492.611937
665718.379520 756796.324489 562460.606255 147740.229453 698537.843876 540532.550586
587595.005615 468844.225446 408502.421221 363662.166111 316690.924505 376806.394304
357425.530815 449914.156127 383980.547441 417882.781716 223374.531713 622735.443545
307603.947296 323102.697878 261938.025796 341196.838406 333626.762014 394538.302550
373649.825240 356705.172767 318045.888846 253912.298600 268243.422284 211362.268487
406750.737659 176988.788119 202948.562954 303495.033523 298810.867558 260138.890868
241695.027102 177945.585896 197177.512939
# Output patern 35:
0.420000
# Input patern 36:
63364635.191112 39131024.567879 24164214.835757 15285450.892739 15391232.682623
9372685.905421 7994348.593778 6263341.572593 4813693.098132 4622021.648423 4085068.089069
2873146.616472 5354484.458549 2448628.427297 2457864.809119 2397286.885549 1693304.919773
1595212.653895 2313971.716067 1399954.846806 1035904.075238 1355391.915535 1565963.722838
1092618.041844 1251188.187475 784040.395898 631084.170705 861866.661256 1126696.060819
636364.785645 703687.790318 757117.447507 640113.133631 644120.303605 605709.331185
801431.010121 561532.282351 515185.696175 526386.778053 519847.938139 628961.540981
763484.847413 464116.227339 713870.292874 501161.534968 519422.792429 412816.191455
739273.277743 389437.709766 570745.660339 456327.211759 534889.567194 590577.222359
490100.246540 459063.600910 367230.197825 274146.735364 417674.891297 359234.927385
456945.352700 377844.848033 443005.605957
# Output patern 36:
0.510000
# Input patern 37:
64362618.975609 32111958.940379 22750917.470780 15462569.033354 10792952.582079
9486206.552043 5728799.596624 5780341.039060 1917250.741215 4765599.688167 4120513.849372
3022285.441811 2350422.146168 2499149.134883 2338373.416747 1762031.760400 2355937.423258
1648282.084469 1902328.145082 1512092.517755 1399527.193692 1129288.195738 1040743.002921
1168283.958492 973952.722799 991418.851077 1226992.507617 1144664.390832 983445.823218
1274175.712402 769668.670180 488338.671258 635952.311136 717127.724978 695699.585661
595775.061059 737513.766874 724168.585136 802284.831907 745997.872420 613199.630062
558476.948765 270779.792195 466913.321461 604119.261912 457882.532109 414153.184958
485349.994747 600772.081762 508323.198301 487701.920269 486338.414971 370029.433145
312450.754973 457920.828980 377974.112673 382507.925355 351346.618552 399907.299157
459908.024167 368712.618160 492436.529785
# Output patern 37:
0.460000
# Input patern 38:
65958562.104817 34379896.042089 22037648.410946 14630389.961120 10057276.265804
8252434.256959 7288812.839707 6297774.544556 4465046.965938 3014924.407860 5562139.255912
2851122.184643 1989805.181643 2611931.046948 2110984.703546 1658356.443939 1491212.822232
2066466.073875 1582100.038659 1275407.526504 1111433.757209 937393.698052 779901.842203
1504629.596304 1040796.028231 1041105.886062 900149.249305 701739.191587 951684.665719
895863.122803 661673.632283 759568.388554 409753.287973 824572.533596 697083.413664
715130.365962 393095.377519 400859.152633 508513.049911 623698.445151 623692.675429
475269.550095 633021.431542 488588.802154 409387.093453 627652.088306 415049.128620
475266.713978 444356.152914 455072.118828 546362.578913 331572.919756 342885.836722
384493.840598 498951.445089 230452.569877 439425.886855 543024.713768 397303.820081
434992.005449 493429.437118 281107.958740

```

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# Output patern 38:
0.550000
# Input patern 39:
57598206.473175 36062178.656572 22767976.669100 16343724.925311 12415176.761460
8855133.175034 7191049.088562 5132515.243464 3826706.615307 3794855.032939 1183306.426217
2655926.154082 3069277.330743 2399771.429282 2483374.701303 2037803.529683 1544538.059803
1584953.953149 1516609.430933 1268734.017304 890840.182447 1230045.757179 686741.170712
829222.032771 544875.018310 997166.656224 1054013.032130 924121.985945 721204.294871
800319.350586 648742.398523 681539.210610 640852.879629 304152.996528 740841.347082
500788.100927 555980.749814 423038.175799 645691.061893 683826.551496 411197.965394
400288.104962 545532.340249 377861.781696 533600.903664 455511.324624 258259.381714
468182.568442 404271.902902 227747.443540 464798.805098 454514.252137 479145.083187
351766.821676 345515.066209 364575.799372 329577.870784 443888.648273 357258.888540
432527.931394 311528.871060 411232.536621
# Output patern 39:
0.490000
# Input patern 40:
56020951.422488 37910037.574439 21396547.338582 14355776.950724 12506911.328189
8417229.395708 8397070.472861 5980576.021779 4231801.462678 4094010.749830 3612279.032389
2874280.633687 3177928.156325 2808258.118164 2536341.743754 3586508.580645 1347587.495008
1517176.884669 1663695.680266 1437486.736003 1488907.322008 1456069.443246 1102069.489724
1044782.746969 652120.943108 1104327.496284 1069140.461256 932303.494211 838336.816414
844383.673828 616262.146649 564543.003056 579710.523348 403083.830455 909445.092356
796134.455440 663801.782540 681176.048118 603077.843541 556058.497223 680139.119974
689812.778097 790206.048674 436296.102160 533538.182891 522336.255860 515620.385227
453421.630277 547411.615948 575679.498762 534251.776725 554411.983310 377035.910480
355645.310193 402806.793057 385603.031229 442395.762302 514580.691370 441487.589424
424674.534042 383805.196565 611771.202637
# Output patern 40:
0.390000
# Input patern 41:
58717006.437423 33685267.699877 23926124.768115 14883239.568135 11828306.179475
11398330.468698 6085481.884472 8421727.526390 4685370.864728 4563744.066995
3884760.231822 3003173.930576 2362915.487617 2808712.481460 1841597.083940 1470894.511456
3272037.867483 1826973.140558 1589444.449314 1284767.668546 1396184.254038 1258728.363274
1310392.418137 989240.023692 827082.475711 1170090.227615 483524.918729 898773.609497
875396.679412 904836.736328 984794.787658 893231.419610 871624.326150 921037.944947
905672.059636 606730.102291 567536.590070 608923.657286 825183.641330 663310.880977
432145.653652 551824.693352 685975.483765 417045.887830 535558.963851 685730.605454
516745.792587 516687.450538 387564.845707 431695.560156 417153.054266 492501.659356
400668.127192 344498.198242 533686.166435 543017.566554 493827.411911 325787.889156
449761.920150 541220.286159 469089.301449 511840.937500
# Output patern 41:
0.440000
# Input patern 42:
63044877.893785 34959219.029924 21589132.803250 13630024.350492 12332743.762753
10643219.558662 10566893.950538 4008891.189209 5742497.457391 5760929.450164
2846294.197539 2975694.128874 2231875.325508 3197734.881219 2492440.839347 2016334.011908
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# Output patern 42:
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# Input patern 43:
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# Output patern 43:
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# Input patern 44:
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# Output patern 44:
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# Input patern 45:
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2014120.893642 2256150.858404 2314394.414022 1335929.645325 413809.262250 1024478.400221
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# Output patern 45:
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# Input patern 46:
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# Output patern 46:
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# Input patern 47:
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# Output patern 47:
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# Input patern 48:
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# Output patern 48:
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# Input patern 49:
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# Output patern 49:
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# Input patern 50:
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# Output patern 50:
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# Input patern 51:
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9330721.444005 6836602.539025 6586755.649739 5565634.557836 4604016.168530 3527481.666383
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397456.541981 623222.839041 524639.947510
# Output patern 51:
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# Input patern 52:
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# Output patern 52:
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# Input patern 53:
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# Output patern 53:
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# Input patern 54:
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433938.818056 385182.313023 504182.165908 457809.371094
# Output patern 54:
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# Input patern 55:
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# Output patern 55:
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# Input patern 56:
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# Output patern 56:
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# Input patern 57:
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# Output patern 57:
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# Input patern 58:
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# Output patern 58:
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# Input patern 59:
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288129.063948 389009.436422 347873.529870 280235.445557
# Output patern 59:
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# Input patern 60:
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# Output patern 60:
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# Input patern 61:
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# Output patern 61:
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# Input patern 62:
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# Output patern 62:
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# Input patern 63:
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# Output patern 63:
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# Output patern 64:
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# Input patern 66:
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# Output patern 66:
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# Output patern 67:
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# Input patern 68:
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# Input patern 69:
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# Output patern 69:
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# Input patern 70:
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# Output patern 70:
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# Input patern 71:
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# Output patern 71:
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# Input patern 72:
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# Output patern 72:
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# Input patern 73:
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# Output patern 73:
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# Input patern 74:
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# Output patern 74:
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# Input patern 75:
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# Output patern 75:
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# Input patern 76:
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# Output patern 76:
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# Input patern 77:
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# Output patern 77:
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# Input patern 78:
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666530.826449 701517.928406 619033.305562 559467.835837 610941.375876 439373.764880
558494.185977 712558.109387 619147.060800 842044.865003 440190.215642 684609.089764
665890.736453 617590.087815 526018.615234
# Output patern 78:
0.520000
# Input patern 79:
79087326.851000 39528132.900498 29863513.670900 17570600.814092 10816464.281837
9645155.575985 7376230.099903 5688321.435045 5172184.063528 4193654.126607 3526200.794421
3433391.491659 2820065.717644 2233590.738825 2171145.800250 2061760.384496 1922138.338999
1900698.581266 1659923.360594 1445627.131155 1428701.833501 1221617.047898 939801.250728
876438.166984 849574.380462 871672.773148 974090.251066 901308.602500 796278.017416
882434.945801 851495.367764 678756.963914 599745.626753 527312.666314 543831.375656
745423.196578 615382.457597 211493.655250 341122.537969 746481.847181 652538.986175
457686.622206 569558.111160 490299.284042 241731.696685 326767.652777 524767.875763
495077.814677 417346.452625 372690.659442 322992.352780 360467.546046 391821.452574
333204.711101 349555.264523 444432.627215 464042.651750 435004.799424 377466.124353
294828.275484 293920.576957 326944.896484
# Output patern 79:
0.510000
# Input patern 80:
60368034.416790 38953443.266763 23621493.611183 15149186.884041 11541430.307213
10055264.837158 8583276.430360 6640060.874208 4736564.607826 3516636.900043
3062863.239936 3011867.072138 2924677.521809 2626449.260745 2240255.565649 1961550.818254
1842535.493627 1748552.465355 1582118.427946 1345061.454119 1145002.957584 1054338.429945
1071513.715763 1103678.051751 1055870.623285 947144.138604 840646.093084 801210.435772
822701.398474 850801.625977 833892.181261 744599.942400 610198.606496 447615.661372
304193.196175 197816.643861 162378.993428 208451.185159 314985.763422 419111.404017
465067.176439 439464.020210 382813.381327 353869.081478 370300.102016 396210.255857
378276.836569 304526.045522 225697.134353 210035.464399 282194.747043 400365.215893
481027.527654 475108.860238 411966.288269 353048.233836 347970.026966 381957.805912
405901.006102 389939.233897 349158.852212 327160.960938
# Output patern 80:
0.500000
# Input patern 81:
80254121.014754 37747082.798227 25204023.962756 16154096.041604 12202641.843441
8790657.013791 7562601.999345 5372960.799063 5255714.471269 4512242.688378 3853143.673886
3208154.930244 2978750.805451 2386800.450870 2292632.373325 2032095.096838 1606577.506252
1379881.295392 1480910.381046 1539545.599383 1334608.835156 1163407.186838 1762442.275276
570986.907230 771560.008017 635066.562228 1023981.503414 1124158.581079 645187.406245
731679.099365 771199.139733 647861.473005 604354.630623 707022.476547 428087.757968
567446.131237 634013.286234 574489.868668 444227.634725 611980.124048 433632.696026
522104.904924 526958.626270 433112.532678 210506.423129 530912.904110 502421.188362
346170.228781 584829.176726 417011.368984 403549.445210 294299.794277 238567.236240
534511.904336 409687.951153 405622.752070 338775.707646 440522.325242 323708.112621
466976.641406 319596.530043 345212.175049
# Output patern 81:
0.470000
# Input patern 82:
54670824.610140 34040285.908463 25141264.994959 18710151.389849 11577558.173126
9740114.396361 6667470.566039 6493605.506866 5276955.046830 4508363.657872 3084652.006010
3246785.305475 3100677.266677 2473890.240715 1881007.472231 2224869.026578 1624423.158626
1753656.449729 1499069.887346 1257878.703027 1256576.671553 1303325.012029 972390.399358
1068693.412254 940999.179761 889660.735880 748387.600138 994816.733648 711353.559234
527580.756592 828972.385731 779072.499644 462646.300134 649093.463305 484259.123114
741864.316031 559356.495613 208332.085638 795125.315181 416466.815855 509864.660975
348759.222416 558610.132244 376741.810975 490286.433634 422310.846687 322485.917267
473488.716297 408535.832298 440565.874110 261358.223511 404926.888244 515878.956557
276029.979605 305749.272390 494468.936261 299935.334105 372943.119865 376230.007458
348891.930061 347878.919699 392513.068604
# Output patern 82:
0.500000
# Input patern 83:

```

```

65357969.458653 39858509.954108 26376872.988788 18034971.187234 11206979.573608
9142649.129602 7505630.599668 6313296.019936 4457624.643512 3651903.219312 3824517.502481
3451733.615441 2715777.292908 3890355.925540 1452697.676369 1522377.607203 1984792.340572
1456505.841457 1399374.656487 1333225.691918 1452208.267664 1263636.506773 952508.966737
1083034.457911 1010322.077765 1089866.943211 625629.019567 688689.125135 601217.505190
853645.571289 725313.489963 727820.970908 684750.847683 616239.158058 585277.047562
419498.650576 586257.367894 475755.979949 582590.215762 362255.142303 515295.990567
464374.551350 544553.745722 481186.555894 520253.220166 178139.327390 478278.112731
424914.639078 256918.514705 489481.697767 427407.230227 475066.831132 264410.255312
386909.971567 416598.417886 407955.138048 306290.358718 256617.990207 432361.050688
330361.488649 295412.793883 589279.011230
# Output patern 83:
0.550000
# Input patern 84:
61789138.500788 30269320.113888 25112276.003052 15115311.939135 9910870.316835
9527967.383837 6182257.887198 6285758.014314 4073435.551784 3899968.209344 3542026.349091
2827852.323501 2562426.380046 2016827.660904 2247344.050270 1756073.686476 1562015.129109
1340309.684965 1628828.042943 1102153.385825 1102102.018464 1045380.770309 1141218.712391
859792.335840 725392.481779 904276.376049 1034370.994557 169614.365884 1188864.974748
392309.165771 736169.140904 494986.463646 680316.852200 480306.579236 544883.196548
556123.988999 359845.404609 641157.239639 362183.452952 433682.866065 477225.350578
407573.915476 439732.174546 303952.777830 500501.090904 346438.809311 375081.935419
309296.882863 491934.339231 258501.630651 379161.882475 330466.539029 394590.759524
325162.736097 251464.225614 457481.393538 300712.827165 241565.699374 462121.346024
223602.841580 401816.447572 264074.569092
# Output patern 84:
0.480000
# Input patern 85:
65988433.737284 37914793.566238 23321630.454952 16868911.880672 11979161.839838
9316346.304170 7383662.840136 5858422.561932 5050697.854674 4054758.115199 3602055.656378
3089209.963717 2597746.069282 2462126.112179 1941315.673961 1942272.938299 1528170.570995
1199906.650414 1332117.125555 2559513.448949 1239825.815028 1389772.234245 1179311.424549
1035606.300268 1068861.146093 882003.533090 928037.679881 817623.166862 777910.075025
785365.822754 675600.519755 731059.483105 628845.949010 638683.716530 639654.637424
542148.178721 662045.571558 491317.639525 654773.227196 745368.211167 597744.830358
191453.990168 505884.175073 355940.744462 415908.750515 422917.630008 363611.412613
433182.079040 355913.246198 399193.966824 380372.084855 354160.958331 404222.713864
331470.845497 392658.357712 352046.628130 344386.670655 404316.891225 279813.448541
457756.167923 235669.230499 131660.692383
# Output patern 85:
0.510000
# Input patern 86:
62798708.816951 37353553.043481 22423562.860099 16784589.261902 13558809.483441
9790653.166388 6881850.786381 6222389.006241 5049548.591365 4355549.017573 3841718.142868
3345482.232864 2524298.079722 2315392.240707 2501749.169910 1906581.202274 2038369.475724
1469469.002703 853511.040042 2501877.205985 759649.863575 1095063.106975 1203041.841555
1018985.283130 1046042.052056 891955.240978 867026.280843 717156.831153 864269.936596
736992.963867 712181.380483 671238.532329 615157.881352 604868.257879 598198.313322
699739.270278 435275.525922 583966.332832 402921.700840 675261.549658 539066.932127
293869.719418 499861.258836 571483.218381 380715.449114 466307.980973 386634.967817
445216.494486 395059.329821 465625.300487 381796.798562 374147.135632 410371.795709
377668.858648 421994.417649 370429.738309 428790.668595 265474.155762 446976.727111
395283.968648 445197.370273 211777.578613
# Output patern 86:
0.480000
# Input patern 87:
39549882.489566 27812278.758118 14935720.811152 14257813.396588 6606688.022369
7931388.627997 7721843.326979 4297802.476952 4421652.968057 2984406.221045 1703454.121202
1468598.700287 1057131.923069 1686303.473057 1109942.193329 850061.863063 1617936.861464
458078.828969 1359380.570351 853234.137242 1383355.233223 891712.169498 790500.173707
815289.068620 438142.562238 436093.831552 602026.898810 529470.768513 505604.323082
488804.998047 458954.036444 511998.309989 308797.925461 650423.216545 408558.466673
381664.676173 328169.445070 238593.131118 254295.331459 182105.222794 368521.540436
348847.449571 270142.752997 380529.948130 307977.785387 343539.586514 302702.430465
411909.851475 294555.524776 253253.321176 163567.963252 229453.694213 74685.790073
214821.956152 207514.746660 231450.375470 181883.087827 213569.560523 275780.883521
209673.610404 335460.698123 324643.816895
# Output patern 87:
0.410000

```



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# Input patern 88:
64424846.733161 34922951.650682 22712194.096095 17672696.823327 12885488.428203
8757046.180862 6997111.120434 5947392.254181 5687386.222610 4433827.990850 4154690.560592
3220610.252665 2955653.963149 2267043.123821 2398958.722692 2434507.313311 2464959.252129
2279192.788864 2030256.276336 1875053.146731 1224309.223308 784681.486326 502425.184542
954043.304589 1150061.477638 1103273.327161 847614.301173 951145.759094 1130666.609407
1103642.493408 920652.307936 841372.563544 779556.521146 679072.455825 565378.153668
716849.145151 746257.849786 683795.987942 476687.437026 566661.196902 646026.936685
651023.406738 638424.773380 752679.808401 839530.011447 618826.171590 462536.999253
503173.345006 768827.187871 735678.576848 573171.457691 450327.830802 535771.385504
566079.297019 496672.562314 457805.880693 497558.314566 507043.773735 419269.636250
374214.547440 430543.753632 482325.032471
# Output patern 88:
0.470000
# Input patern 89:
75701384.805786 36521659.564531 25004522.319198 18332473.484283 17797462.083000
9216327.758089 4802032.478215 4780670.859831 5487311.664204 3793434.205970 6582259.310946
4579964.801309 3239693.408418 3662644.135698 2171399.031323 2582515.337524 1160898.774170
2171653.572520 2266235.624976 2187792.498179 1865884.565011 1586530.322949 1103875.850448
1631025.087963 1690695.625942 832234.474095 861661.213723 1227728.460185 1122868.166303
1292903.649658 1130369.520668 898978.012133 883634.933203 966494.013323 788020.397595
745864.155753 773839.257057 815016.790912 733998.910930 819296.168724 775317.809825
845340.473011 847965.955148 1015094.100100 627424.073725 608837.044478 556971.957775
563530.856041 633875.388834 705701.487750 562632.934291 665209.062124 851003.019545
514693.736448 635260.449332 436829.259098 506719.240644 568740.851378 545455.245695
428935.870179 634583.695139 671820.517334
# Output patern 89:
0.460000
# Input patern 90:
56873823.678541 36460705.272242 22441657.923858 15213109.448931 10818100.887090
7659692.497592 5984621.987306 5680170.168866 5583782.474689 5569999.825846 3529408.194899
2525390.915147 2561166.824167 2226423.870216 1580626.262988 1638523.424652 1882967.654598
1563875.336668 1478992.508070 1727540.812269 596359.739696 1424164.460473 1066720.134445
1152399.001768 959096.397628 865749.370049 988560.728762 856189.859745 627771.746078
652934.710938 637347.928381 442711.917911 652645.351336 373606.668399 337734.091026
451919.254756 533586.277968 409174.735585 371133.298754 460862.274593 397526.844366
341641.630656 452635.352028 267994.736313 532853.032862 383616.799706 478870.953774
326154.204863 218295.058064 340885.303422 333841.038098 234931.248441 354335.270250
445450.900100 300397.835817 434255.133171 243553.016298 293213.506780 320830.884963
433756.374152 443642.237632 393031.202148
# Output patern 90:
0.460000
# Input patern 91:
68036734.530020 37491469.625629 26521955.031326 17196037.109250 11286388.918006
8875403.226225 7554276.732708 5801510.112327 4964007.947189 4203238.835549 3726297.102849
4003173.554609 3356835.765694 1194233.900412 2335622.377632 1820325.611984 1825463.686462
1731385.602132 1786525.145481 1194996.398993 1251120.890219 1240190.296245 1073986.500907
830000.312899 1027627.263487 764102.596827 928626.812794 813264.878626 891744.612619
915293.184570 332390.720513 648726.022382 580428.367155 731702.671659 573574.752545
501063.253072 642707.008234 551068.613284 426511.176938 485607.441879 516857.381521
356013.739823 499215.681800 334303.448199 554202.379276 639487.387674 182292.263406
429754.803525 339861.637395 467307.766418 283222.636269 449422.415477 436191.673379
350393.418934 334287.526710 466893.474936 304115.503126 338626.298682 346022.479741
311223.322383 350779.376736 524598.623535
# Output patern 91:
0.490000
# Input patern 92:
63104664.799809 38026927.060520 26036433.676659 15315728.646144 11446596.019950
9880581.835451 7481459.482564 6528397.015786 5641748.851612 4278482.015120 3607670.558269
3086039.217830 2744858.085367 2848820.517432 2260128.811718 1607584.528952 1937931.857414
1812396.887090 840565.113669 705881.476174 1147682.199247 1092872.941226 1002160.385920
973127.035050 683178.290304 614303.321812 826535.289623 852230.306080 801318.686268
849530.553711 843752.869016 823093.829244 774364.407950 673170.007867 691843.048613
600512.948842 318634.075020 351168.517995 619221.763838 605964.137768 512462.935575
544621.410255 466386.105168 403821.942369 484078.597704 429630.182763 325305.091520
389014.701831 436373.145007 378584.624895 365550.875346 397408.745657 459364.662510
502699.945562 446669.523591 441205.157670 476997.346425 338432.738847 229544.746624
370302.751891 473318.750524 457939.869141
# Output patern 92:

```

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0.510000
# Input patern 93:
50778649.493572 31800159.718605 19540575.577791 15224908.240709 10519520.177881
8764149.292025 6575378.211495 6220961.522424 4437552.241863 4247846.021867 3198954.066623
3276431.341555 2425558.260648 1859964.107490 1961364.906050 536712.731817 1619054.496426
1967589.912959 1356190.116425 1076206.443286 1105844.627269 1033400.723746 882784.991061
811292.963135 709681.205516 870270.672329 612257.871380 702091.781131 585686.190426
720287.711182 604053.622767 399136.388552 644807.998843 428470.862544 661346.322858
841934.884782 633744.725387 447894.342647 560990.221017 534858.103891 463935.976195
435048.817538 378495.551117 493027.604719 326687.406845 373054.467217 320674.330500
390658.105114 351269.996085 192085.564539 393814.239375 383265.116718 421450.452777
498754.832013 420105.159502 251115.904394 393882.091734 403748.818291 356941.249518
372716.090885 330883.200460 459426.382080
# Output patern 93:
0.490000
# Input patern 94:
59314651.879962 32386800.816927 20383380.511271 15540654.615261 12509634.441732
8149197.279485 6531019.050334 5547556.285143 4486671.194196 4115127.556008 3485153.868143
2752358.908843 2173632.076118 2677935.185519 1633242.154204 2268735.012906 1559443.262453
585824.104545 2446407.200412 834172.448296 1142341.711315 923806.754860 1115362.397033
990291.975145 830609.877688 750195.278287 773835.221874 739266.601648 717424.893527
653322.546875 604189.134586 569472.771226 549046.479375 681782.805776 358376.820870
696206.737441 268068.030903 533047.257658 652655.635616 265364.430549 419559.680398
444278.951196 481752.844625 390151.185102 294220.451068 459511.336942 341219.168281
411779.471955 352794.730500 330294.132934 358459.411452 347039.061069 367249.845278
357533.721669 258558.023222 425514.874169 208280.186582 476614.940264 293485.431551
247023.431382 345747.192829 392301.601563
# Output patern 94:
0.490000
# Input patern 95:
79545648.139703 41700863.869828 24719483.515273 18338410.233836 12223081.894069
9383578.808253 7514350.722357 5415573.209590 4864313.391783 4191429.080455 3450356.521160
3330209.934151 2629781.898554 2395502.322576 2953561.992774 955536.855699 2311039.784768
2036221.411554 1763793.301983 1058461.617326 925062.031767 960442.177227 983153.147799
1137317.779881 1062052.175316 951959.247309 949002.066450 831500.076783 793694.886926
719883.368164 533794.717509 620494.806018 662970.163338 552651.299760 448463.546581
778221.111290 571056.558276 413961.631988 477797.126498 508215.625114 571976.043266
564939.626762 421222.974108 357930.580986 336816.190237 356220.685237 447940.492510
428173.439423 405474.998520 456275.395735 414626.143860 541858.847918 136567.262151
522648.952844 297848.189249 288142.324390 329029.450559 366927.095940 448131.306573
408075.377814 331014.293805 328237.996582
# Output patern 95:
0.480000
# Input patern 96:
59287048.012957 37985727.409748 23781618.703728 15504905.870205 11930131.013989
9749846.633046 7123726.896567 5603133.774041 4989094.654182 3229194.379991 2976299.577330
2445595.437762 4411964.537876 1866112.151249 2664247.572213 1811619.477586 2043411.340546
1791241.511378 1485155.474838 1472446.956496 1226113.345714 1183871.498052 971516.999716
987486.342200 1360707.256761 929940.022715 1245122.843748 695161.535649 815295.081837
869497.177734 928828.710590 450923.100491 776188.829059 640496.197104 490895.820657
646938.002614 632284.641536 673244.080865 645121.852229 458664.740686 593480.527587
304780.886949 635294.838055 482220.445440 301779.481902 390373.218868 532234.098804
279133.205954 491375.661567 472974.930996 423221.132501 399813.424477 348376.669796
285049.969678 403496.490937 339751.014729 385878.063447 221382.454087 358455.465377
359286.814724 411431.833220 345087.184082
# Output patern 96:
0.490000
# Input patern 97:
62921958.492717 33358238.919052 24043743.303205 15301487.967960 12556004.425264
9309002.460529 7921387.486309 6044017.256219 5601494.050303 4359522.082533 3792743.423222
3287593.057404 2914462.245827 2082156.188472 2038949.184896 1331518.340947 3005242.187056
1835488.204302 1433482.580109 1466050.605644 1263652.547883 1071685.971272 1043711.909897
1030771.776094 840798.150682 889951.328271 848069.092602 787876.266593 716812.014947
856019.295898 675526.027809 688417.298657 838218.651154 818207.235527 963759.633441
495007.582600 493872.408213 670835.838650 474656.516027 526032.546984 488684.284314
519685.217492 375262.389730 504854.545145 385191.163949 411189.594731 360861.570092
465241.915696 280641.224824 429249.473312 447474.233724 331257.681193 281666.772128
167113.282284 312525.900097 436886.216337 257857.055317 410818.449668 354701.132914
397292.121404 312764.107254 456233.678711

```

```

# Output patern 97:
0.470000
# Input patern 98:
63033740.112103 39365472.378977 21548407.244453 16420247.022219 11973420.652289
8600504.328098 7534092.131372 5607118.829085 4675471.706626 3707011.281557 3931666.857081
2351529.005349 1837091.248358 1285886.094135 2725021.943056 1766010.317771 1819593.209754
1703824.969523 1422221.552814 1116163.959357 1335915.505637 1059165.803021 1002386.258824
901407.284235 800313.993620 850886.037512 806240.877044 460535.971371 422169.917776
918566.564941 593262.975597 722637.618439 733834.024591 445361.203983 633895.077175
378691.511328 602173.738023 474377.764549 338214.041844 516580.022086 342399.410232
510141.968142 268959.802212 314804.868424 407470.121374 547405.324810 325245.781654
525493.989377 382599.081602 241965.774535 405514.205129 326055.738318 343310.607547
340830.784020 234106.409259 404006.351315 297115.652415 330275.702177 205934.361608
401993.775530 377169.216762 392375.640137
# Output patern 98:
0.490000
# Input patern 99:
68792713.861441 36716213.804654 26906229.956675 16243926.149319 13558615.719212
9164483.420224 8013581.934469 6198855.920759 5082612.546581 4292872.768058 4233519.201594
3006213.408948 3321602.142513 2592407.216251 2706440.920518 1726794.884370 2343304.408584
1349509.026734 2415772.035833 1889033.009711 1067524.724312 793394.087925 1086902.817354
952067.976739 817221.431843 875114.090642 1009155.609761 758160.248916 850519.301468
786055.243408 767081.393679 610852.837221 683746.118986 523451.948760 752305.319095
338985.820699 782860.885784 629907.667456 561607.852789 599045.142457 421109.138831
518390.593226 841949.970608 575138.660948 536349.359171 554435.744198 499660.431544
431680.769288 418675.777721 396496.258244 513650.668674 296316.914806 532366.343550
392001.092950 478671.400233 331596.312920 430861.370998 373163.503823 211411.381661
261455.814710 401363.501907 327118.600342
# Output patern 99:
0.480000
# Input patern 100:
64373935.571038 36203648.606389 22045415.461697 15965280.310038 11208352.182805
8874774.668761 7380684.049138 6331903.018380 4261484.190872 4999343.364022 4456189.449805
2280136.396013 2639689.795858 2721403.399874 2551870.731824 1682026.602919 1696704.224759
1703871.534163 1452413.408416 1754219.082781 1981818.573859 1476761.813219 1219910.399920
1042461.298815 1285941.138642 801057.997120 807021.532588 959864.171875 1107551.871191
571019.359619 1000241.851318 524842.329864 690533.178228 567274.456751 827144.734136
777111.326534 418226.787921 689714.417472 660944.232680 609669.087008 545902.110563
532882.526650 422029.690787 383224.037310 337598.562069 454667.357450 375555.439289
330483.501268 584309.081538 459310.752330 272657.742648 405186.464758 306050.322706
452863.695275 464086.361225 385130.933573 503709.050633 254781.432497 427644.951635
335191.588227 400818.675347 210670.287354
# Output patern 100:
0.550000

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APPENDIX C

NEURAL NETWORK SAMPLE TRAINING AND TEST SETS

Table C.1: Neural Network Training Set.

filename	mpd	etd	avg-std	overall-std
6091403.11	8.007	6.605	2.611	2.676
6081442.35	9.740	7.992	2.254	3.059
6100953.29	8.341	6.873	2.178	2.995
6091401.54	7.871	6.497	2.530	2.614
6091137.25	10.792	8.834	2.316	2.698
6211311.25	6.274	5.219	2.232	2.353
2091102.16	9.728	7.982	2.791	2.945
6211142.06	7.886	6.509	2.867	2.963
6211351.05	7.162	5.929	2.191	2.269
6091628.14	6.609	5.487	2.053	2.113
6091128.55	8.277	6.822	2.575	2.638
6091632.41	8.895	7.316	2.899	2.971
2091056.08	5.917	4.934	1.773	1.886
6211248.04	5.104	4.283	1.847	1.891
6091303.00	8.020	6.616	2.040	2.227
2091245.59	6.501	5.401	1.981	2.054
2091244.16	7.287	6.03	2.308	2.38
6091355.42	7.322	6.058	2.334	2.421
2091102.31	5.636	4.709	1.643	1.824
6211349.47	6.278	5.222	1.866	1.900
6091406.26	7.431	6.145	2.489	2.545
2091103.59	9.01	7.408	2.58	2.757
6101104.03	6.915	5.732	1.939	2.018
2091103.52	9.464	7.771	2.675	2.827
2091240.51	5.198	4.358	1.58	1.687
2091245.37	6.404	5.323	2.05	2.129
2091104.06	8.442	6.954	2.394	2.575
2091055.45	8.790	7.232	2.498	2.731
2091054.56	7.085	5.868	2.053	2.221
2091241.50	6.748	5.598	2.191	2.26
6101513.34	5.857	4.886	1.608	1.672
2091245.44	4.855	4.084	1.521	1.716
2091241.06	6.101	5.08	1.933	1.997
6091633.56	8.813	7.251	2.874	2.942
6091405.04	6.716	5.573	2.087	2.146
2091104.21	8.384	6.907	2.405	2.544
2091103.00	7.075	5.860	2.021	2.193
6091358.11	7.140	5.912	2.385	2.470
2091240.58	5.881	4.905	1.827	1.905
2091243.18	5.527	4.621	1.763	1.819

2091242.42	5.42	4.536	1.747	1.824
2091106.05	5.036	4.229	1.505	1.682
2091105.48	7.068	5.855	2.005	2.147
6081126.05	10.302	8.442	2.095	2.693
6211206.52	6.985	5.788	1.365	1.492
2091211.07	5.226	4.381	1.572	1.64
2091053.29	6.095	5.076	1.776	1.903
6091400.36	7.893	6.515	2.565	2.636
2091209.02	4.145	3.516	1.28	1.327
2091230.12	4.187	3.55	1.292	1.341
6211340.37	6.684	5.547	2.023	2.085
6211041.17	7.691	6.353	1.656	1.704
2091107.18	5.65	4.72	1.638	1.8
2091232.31	4.492	3.793	1.394	1.449
2091232.39	4.467	3.774	1.414	1.47
2091055.31	4.069	3.455	1.219	1.316
2091232.16	4.327	3.661	1.345	1.409
2091053.01	4.979	4.183	1.503	1.638
6081446.50	8.374	6.899	2.155	2.323
2091230.43	4.42	3.736	1.362	1.412
2091210.30	4.585	3.868	1.369	1.439
2091057.20	4.985	4.188	1.487	1.741
6081355.03	5.953	4.962	1.859	1.917
2091323.07	3.37	2.896	1.066	1.103
6101427.44	6.227	5.182	1.640	1.702
6091133.44	7.455	6.164	1.820	1.940
2091056.38	3.661	3.129	1.107	1.191
6100934.54	4.625	3.900	1.225	1.272
2091321.21	2.983	2.587	0.946	0.968
6101322.18	3.360	2.888	0.801	0.835
2091324.29	2.783	2.427	0.866	0.891
2091322.59	3.474	2.979	1.082	1.105
2091155.10	2.15	1.92	0.687	0.719
2091323.22	3.065	2.652	0.947	0.966
2091154.48	2.195	1.956	0.694	0.717
6101037.24	2.875	2.500	0.760	0.769
2091322.52	3.059	2.647	0.945	0.962
6101134.22	5.640	4.712	1.479	1.599
2091323.51	2.594	2.275	0.777	0.796
6101533.53	4.761	4.009	1.377	1.414
2091131.04	1.687	1.549	0.547	0.554
6091339.28	5.385	4.508	1.523	1.572
2091158.25	2.056	1.845	0.622	0.641
2091155.38	1.835	1.668	0.593	0.619
6211318.11	3.682	3.146	1.301	1.513
2091322.31	2.296	2.037	0.707	0.729
2091129.30	1.745	1.596	0.541	0.564

6100948.37	3.788	3.231	1.016	1.045
2091155.03	1.825	1.66	0.577	0.595
2091324.38	2.47	2.176	0.759	0.771
2091325.00	3.147	2.718	0.967	1.042
2091156.50	1.542	1.434	0.49	0.498
6091356.22	5.598	4.678	1.763	1.908
2091134.00	1.594	1.475	0.506	0.518
6091623.01	5.090	4.272	1.443	1.477
2091323.29	2.361	2.089	0.726	0.737
2091326.11	2.547	2.237	0.796	0.805
2091322.21	1.922	1.737	0.601	0.608
2091324.14	2.607	2.286	0.846	0.913
2091134.07	1.381	1.304	0.444	0.45

Table C.2: Neural Network Test Set.

filename	mpd	etd	avg-std	overall-std
2091242.05	7.327	6.062	2.348	2.411
2091056.30	8.065	6.652	2.352	2.515
2091244.01	6.776	5.621	2.175	2.254
2091056.01	8.657	7.125	2.549	2.741
2091245.59	6.501	5.401	1.981	2.054
2091103.44	11.2	9.16	3.196	3.405
2091102.31	5.636	4.709	1.643	1.824
2091209.21	4.261	3.609	1.27	1.312
2091056.44	9.852	8.081	2.822	3.056
2091056.57	9.051	7.441	2.658	3.052
2091057.06	7.331	6.064	2.202	2.396
2091243.11	6.945	5.756	2.258	2.335
2091244.54	7.029	5.823	2.246	2.31
2091245.31	6.963	5.771	2.222	2.29
2091245.24	7.048	5.838	2.2	2.272
2091104.55	7.033	5.826	2.103	2.253
2091243.48	6.88	5.704	2.162	2.226
2091245.02	6.344	5.275	1.993	2.146
2091102.38	6.869	5.695	1.972	2.107
2091055.38	5.405	4.524	1.625	1.758
2091055.19	5.215	4.372	1.588	1.756
2091055.54	9.445	7.756	2.744	2.936
2091104.49	8.428	6.943	2.539	2.853
2091103.13	9.065	7.452	2.597	2.762
2091055.45	8.790	7.232	2.498	2.731
2091053.15	7.839	6.471	2.261	2.405
2091105.26	7.454	6.163	2.157	2.316

2091244.09	6.928	5.742	2.22	2.29
2091243.04	6.58	5.464	2.142	2.278
2091245.16	6.724	5.579	2.169	2.24
2091245.08	6.67	5.536	2.159	2.235
2091244.47	6.69	5.552	2.139	2.204
2091241.43	6.392	5.314	2.058	2.135
2091245.37	6.404	5.323	2.05	2.129
2091240.51	5.198	4.358	1.58	1.687
2091243.25	4.849	4.079	1.486	1.613
2091056.51	10.348	8.479	3.032	3.256
2091104.28	8.768	7.214	2.516	2.673
2091242.28	7.156	5.925	2.342	2.405
2091241.57	6.765	5.612	2.149	2.215
2091241.13	6.35	5.28	2.023	2.103
2091244.39	6.401	5.321	2.016	2.078
2091057.13	6.028	5.023	1.787	1.919
2091105.03	5.242	4.394	1.537	1.789
2091242.13	4.762	4.01	1.461	1.523
2091103.06	10.264	8.411	2.979	3.224
2091103.35	8.349	6.879	2.453	2.640
2091105.34	7.863	6.491	2.267	2.441
2091056.23	7.103	5.882	2.048	2.198
2091242.21	6.602	5.482	2.12	2.197
2091103.28	7.000	5.800	1.989	2.135
2091243.41	6.379	5.303	2.031	2.13
2091243.54	6.197	5.158	1.998	2.115
2091053.23	6.573	5.458	1.909	2.053
2091054.12	6.135	5.108	1.781	2.021
2091105.18	4.965	4.172	1.468	1.668
2091106.25	9.829	8.063	2.857	3.12
2091053.51	9.720	7.976	2.738	3.003
2091053.37	9.099	7.479	2.642	2.849
2091241.28	6.378	5.303	2.061	2.127
2091241.35	6.144	5.115	1.976	2.035
2091241.20	6.077	5.062	1.942	2.015
2091102.23	6.009	5.007	1.919	1.996
2091102.45	5.556	4.645	1.631	1.773
2091229.52	4.606	3.885	1.44	1.485
2091231.42	4.539	3.831	1.403	1.44
2091231.22	4.262	3.61	1.303	1.356
2091054.20	9.596	7.877	2.774	3.115
2091104.35	8.99	7.392	2.539	2.715
2091055.04	7.847	6.478	2.317	2.473
2091054.27	6.223	5.178	1.849	2.156
2091107.25	6.696	5.557	1.904	2.102
2091055.10	6.531	5.425	1.922	2.042
2091105.41	5.221	4.376	1.827	2.024

2091242.49	5.549	4.639	1.771	1.819
2091056.15	5.084	4.268	1.542	1.646
2091105.10	5.275	4.42	1.557	1.64
2091230.29	4.667	3.934	1.439	1.506
2091231.01	4.46	3.768	1.322	1.389
2091212.49	4.234	3.587	1.32	1.359
2091212.05	4.131	3.505	1.285	1.319
2091053.44	7.301	6.041	2.136	2.347
2091054.50	7.360	6.088	2.076	2.262
2091053.08	6.704	5.563	1.978	2.202
2091106.33	5.601	4.681	1.611	1.776
2091210.53	5.132	4.306	1.577	1.641
2091211.36	4.952	4.161	1.52	1.57
2091211.00	4.963	4.171	1.516	1.565
2091208.33	4.837	4.069	1.48	1.538
2091211.14	4.83	4.064	1.471	1.521
2091230.51	4.634	3.907	1.386	1.433
2091242.34	6.518	5.414	2.143	2.215
2091105.56	5.129	4.303	1.548	1.641
2091211.43	5.001	4.201	1.553	1.6
2091054.42	5.041	4.232	1.462	1.589
2091208.19	5.011	4.209	1.525	1.574
2091208.26	4.913	4.13	1.49	1.556
2091208.41	4.825	4.06	1.459	1.513
2091230.36	4.761	4.009	1.454	1.505
2091102.52	4.226	3.581	1.309	1.478
2091210.39	4.456	3.765	1.381	1.444
2091231.31	4.552	3.841	1.37	1.438
2091229.00	4.151	3.521	1.321	1.372
2091209.41	4.218	3.574	1.288	1.344
2091052.47	5.407	4.525	1.612	1.889
2091106.18	5.614	4.692	1.65	1.765
2091211.21	5.403	4.523	1.612	1.677
2091054.06	5.201	4.361	1.549	1.673
2091208.55	4.4	3.72	1.38	1.451
2091229.30	4.419	3.735	1.371	1.421
2091212.12	4.323	3.658	1.34	1.39
2091106.11	3.964	3.371	1.248	1.387
2091228.31	4.181	3.545	1.275	1.326
2091229.36	4.016	3.413	1.255	1.294
2091229.08	3.949	3.359	1.214	1.252
2091211.28	4.903	4.122	1.508	1.566
2091210.46	4.724	3.979	1.49	1.562
2091230.05	4.473	3.779	1.399	1.444
2091103.20	4.544	3.835	1.332	1.427
2091232.02	4.279	3.623	1.342	1.393
2091229.58	4.355	3.684	1.327	1.378

2091228.24	4.188	3.55	1.283	1.32
2091211.57	4.132	3.506	1.281	1.313
2091210.23	5.347	4.478	1.603	1.693
2091054.35	5.318	4.455	1.526	1.660
2091232.09	4.89	4.112	1.506	1.565
2091232.25	4.604	3.883	1.441	1.494
2091208.48	4.266	3.612	1.329	1.374
2091230.20	4.152	3.522	1.295	1.344
2091231.53	3.976	3.381	1.222	1.279
2091228.10	3.918	3.334	1.23	1.277
2091228.40	3.951	3.361	1.229	1.272
2091210.16	4.007	3.406	1.224	1.27
2091106.57	4.27	3.616	1.244	1.314
2091210.09	4.121	3.496	1.244	1.291
2091243.34	3.494	2.995	1.052	1.277
2091227.56	3.959	3.367	1.219	1.26
2091231.11	3.988	3.39	1.19	1.237
2091104.41	5.361	4.489	1.555	1.667
2091211.51	3.566	3.052	1.136	1.177
2091156.27	2.441	2.153	0.797	0.811
2091053.58	6.283	5.227	1.893	2.106
2091228.17	3.821	3.256	1.166	1.221
2091322.44	3.315	2.852	1.03	1.06
2091156.06	2.203	1.962	0.699	0.717
2091130.13	2.129	1.903	0.675	0.696
2091104.13	4.861	4.089	1.395	1.68
2091106.40	4.588	3.87	1.379	1.573
2091052.55	3.445	2.956	1.038	1.122
2091321.43	2.473	2.178	0.773	0.792
2091133.38	1.882	1.706	0.609	0.621
2091133.17	1.852	1.682	0.601	0.618
2091154.10	2.115	1.892	0.657	0.684
2091132.26	2.057	1.846	0.652	0.667
2091157.49	1.902	1.721	0.62	0.636
2091154.03	2.414	2.131	0.75	0.769
2091155.45	2.105	1.884	0.661	0.68
2091133.09	1.922	1.738	0.626	0.641
2091132.18	1.917	1.734	0.614	0.628
2091156.12	2.054	1.843	0.64	0.652
2091131.18	1.825	1.66	0.591	0.606
2091321.00	3.056	2.645	0.968	0.987
2091155.31	1.987	1.79	0.639	0.652
2091321.14	2.778	2.423	0.861	0.878
2091155.52	1.873	1.698	0.607	0.617
2091132.56	1.735	1.588	0.556	0.566
2091130.41	1.56	1.448	0.501	0.509
2091323.44	3.296	2.837	0.966	1.01

2091131.25	1.8	1.64	0.595	0.608
2091324.08	2.25	2	0.726	0.75
2091132.11	1.906	1.724	0.613	0.63
2091132.04	1.598	1.478	0.521	0.531
2091323.36	2.949	2.559	0.894	0.909
2091133.52	2.365	2.092	0.67	0.735
2091157.56	1.99	1.792	0.656	0.678
2091132.34	1.749	1.599	0.572	0.59
2091130.04	1.75	1.6	0.556	0.574
2091131.11	1.498	1.398	0.494	0.502
2091157.27	1.539	1.431	0.506	0.51
2091325.58	2.963	2.57	0.903	0.919
2091131.33	1.758	1.607	0.58	0.593
2091157.42	2.233	1.986	0.703	0.724
2091157.35	1.933	1.746	0.623	0.636
2091131.39	1.673	1.538	0.535	0.547
2091129.37	1.578	1.463	0.513	0.523
2091130.57	1.559	1.447	0.505	0.512
2091325.43	2.828	2.463	0.843	0.854
2091325.29	2.591	2.273	0.803	0.815
2091321.29	1.93	1.744	0.606	0.62
2091134.13	1.524	1.419	0.493	0.5
2091325.50	2.85	2.48	0.897	0.911
2091321.08	2.478	2.182	0.743	0.871
2091325.23	2.596	2.277	0.796	0.809
2091326.05	2.535	2.228	0.779	0.792
2091324.00	2.264	2.011	0.695	0.71
2091158.18	1.643	1.515	0.54	0.551
2091324.53	2.425	2.14	0.74	0.761
2091325.36	2.435	2.148	0.747	0.759
2091325.07	2.701	2.361	0.82	0.831
2091324.45	2.333	2.067	0.702	0.721
2091321.59	2.121	1.897	0.667	0.673
2091321.36	1.916	1.733	0.615	0.624
2091325.14	2.434	2.147	0.73	0.741

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BIOGRAPHICAL INFORMATION

John Michael Kebrle was born in Dallas, Texas in 1966. He received his BA in Physics from Washington University and then worked two full-time technical jobs before resuming his college studies and received his BS in Mechanical Engineering from the University of Texas at Austin. He then began work at Bird Life Design and began his graduate studies at The University of Texas at Arlington. He has also worked at the Automation and Robotics Research Institute, Bell Helicopter, Vought Aircraft, and Raytheon Aircraft Corporation (now Hawker BeechCraft). Simultaneously, he instructed industry targeted Computer Aided Design classes and later taught sophomore, junior, and senior level design classes at night in Mechanical Engineering. He also worked at his father's stained glass studio painting and etching glass. While working full-time in the medical, robotics, and aerospace industries, and teaching, he took classes at night and on-line at the University of Texas at Arlington, and received his MS in Mechanical Engineering, his PhD in Mechanical Engineering, and is now completing his MS in Computer Science and Engineering. He currently works as a consultant in the aerospace industry, primarily in Tulsa, and travels to Dallas, Wichita, and Seattle, performing contractual structural analysis of aircraft components and systems. He has performed structural static and dynamic analysis and testing, acoustic design and testing, acoustic fatigue analysis, defective part analysis, implemented modern manufacturing systems, and performed manufacturing and quality analysis on aircraft and tooling including the Boeing

737, 777, 787, the GulfStream G-IV, and G-V, the Bell Helicopter Models 206, 407, 222, 230, 430, 412, 427, 429, AH1, UH1, the 609 and V-22 Tilt-Rotor aircraft, the experimental MAPL tail fan and prototype aircraft, and the Raytheon AirForce/Navy JPATS, T-6 family of aircraft. He has been active in music starting at 3; performed his first violin solo at 6 at SMU, and also played guitar in numerous St. Louis, Austin, and Dallas area bands. He has also, published sixteen technical conference and journal papers.