

**STOCHASTIC MODELING OF FUTURE HIGHWAY
MAINTENANCE COSTS FOR FLEXIBLE TYPE
HIGHWAY PAVEMENT CONSTRUCTION PROJECTS**

A Thesis

by

YOO HYUN KIM

Submitted to the Office of Graduate Studies of
Texas A & M University
in partial fulfillment of the requirements for the degree of

MASTER OF SCIENCE

May 2012

Major Subject: Construction Management

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Approved by:

Chair of Committee,	Kunhee Choi
Committee Members,	Sarel Lavy
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ABSTRACT

Stochastic Modeling of Future Highway Maintenance Costs for Flexible Type

Highway Pavement Construction Projects.

(May 2012)

Yoo Hyun Kim, B.A., Hongik University

Chair of Advisory Committee: Dr. Kunhee Choi

The transportation infrastructure systems in the United States were built between the 50's and 80's, with 20 years design life. As most of them already exceeded their original life expectancy, state transportation agencies (STAs) are now under increased needs to rebuild deteriorated transportation networks. For major highway maintenance projects, a federal rule enforces to perform a life-cycle cost analysis (LCCA).

The lack of analytical methods for LCCA creates many challenges of STAs to comply with the rule. To address these critical issues, this study aims at developing a new methodology for quantifying the future maintenance cost to assist STAs in performing a LCCA. The major objectives of this research are twofold: 1) identify the critical factors that affect pavement performances; 2) develop a stochastic model that predicts future maintenance costs of flexible-type pavement in Texas.

The study data were gathered through the Pavement Management Information System (PMIS) containing more than 190,000 highway sections in Texas. These data were then grouped by critical performance-driven factor which was identified by K-means cluster analysis. Many factors were evaluated to identify the most critical factors that affect pavement maintenance need. With these data, a series of regression analyses were carried out to develop predictive models. Lastly, a validation study with PRESS statistics was conducted to evaluate reliability of the model. The research results reveal that three factors, annual average temperature, annual precipitation, and pavement age, were the most critical factors under very low traffic volume conditions.

This research effort was the first of its kind undertaken in this subject. The maintenance cost lookup tables and stochastic model will assist STAs in carrying out a LCCA, with the reliable estimation of maintenance costs. This research also provides the research community with the first view and systematic estimation method that STAs can use to determine long-term maintenance costs in estimating life-cycle costs. It will reduce the agency's expenses in the time and effort required for conducting a LCCA. Estimating long-term maintenance cost is a core component of the LCCA. Therefore, methods developed from this project have the great potential to improve the accuracy of LCCA.

DEDICATION

Send my unlimited affection to my family

Kwangwon Kim

Haesook Park

Kahye Oh

Jungyoon Kim

Jungjin Kim

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NOMENCLATURE

AADT	Annual Average Daily Traffic
AASHTO	American Association of State Highway and
ACP	Asphalt Concrete Pavement
ADT	Average Daily Traffic
ANOVA	Analysis of Variance
ARRA Act	American Recovery and Reinvestment Act
Caltrans	California Department of Transportation
CRCP	Continuous Reinforced Concrete Pavement
JCP	Jointed Concrete Pavement
FHWA	Federal Highway Administration
HMA	Hot Mixed Asphalt
IH	Interstate Highway
IRI	International Roughness Index
LCCA	Life Cycle Cost Analysis
NHS	National Highway System
PCI	Pavement Condition Index
PSI	Pavement Serviceability Index
RSL	Remaining Service Life
SH	State Highway (or SR: State Route)
STAs	State Transportation Agencies

TxDOT

Texas Department of Transportation

DEFINITION OF TERMS

Lane-mile	The total length of all lanes calculated as centerline mile multiplied by the number of lanes
Routine Maintenance	Defined in “Maintenance Management Manual” issued by TxDOT in 2008. Include pavement repair, crack seal, seal coat, level-ups, light overlay less than 2 inches and additional base. Normally, contract amount should not exceed \$300,000 per project. (Table 3)
Preventive Maintenance	Defined in PMIS. Include crack seal and surface seal. In case of thin asphalt, cost is less than \$8,000 per lane-mile and in case of med or thick asphalt, cost is less than \$10,000 per lane-mile (Table 4).

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1 INTRODUCTION

1.1 Needs for Highway Infrastructure Recovery

During the construction boom between the 1950's and 1980's, most transportation infrastructure systems were established with a 20-year life expectancy (FHWA 2002; Lee and Ibbs 2005). After the 2000's, these highways have created severe public concerns with regard to safety and regional economic problems. Since most highways in the US have already exceeded their design life expectancy, a huge amount of money is necessary to restore the nationwide highway infrastructure. However, the recent economic recession has created a poor financial status in many state governments which has made it impossible to increase governmental expenditures for infrastructure recovery projects. Under this situation, the *American Recovery and Reinvestment Act* (ARRA Act) was enacted in February 2009 by the Obama Administration and the US Congress which has included many projects to improve poor systems and plan for large scale investments. The major points for improving highway infrastructure can be summarized as investment of a huge amount of federal money into the infrastructures, up to \$80 billion, to promote the recovery of highways from their present status and stimulate the regional and national economies through federal investment.

This thesis follows the style of the *Journal of Construction Engineering and Management*.

1.2 Needs of LCCA for Transportation Projects

The Federal Highway Administration (FHWA) strongly recommends using Life-Cycle Cost Analysis (LCCA) for most highway projects for new projects and 4R projects: restoration, resurfacing, rehabilitation and reconstruction (FHWA 2002). LCCA methodologies are a comprehensive tool for decision making by comparing cost efficiency between design alternatives (FHWA 2002). To analyze life-time cost for infrastructure, maintenance cost data is the most essential of many cost factors since one-time cost is only a small part of the total cost for a life time. Generally, factors such as initial cost, operating cost, maintenance and replacement cost and salvage value are included in LCCA. Especially, maintenance and rehabilitation projects (M&R) comprise a large proportion in early step of projects (FHWA 2002). Therefore, exact data for M&R costs are essential to produce more exact LCCA for many design alternatives and to make an efficient decision.

2 RESEARCH SCOPE AND SIGNIFICANCE

2.1 Gaps in Existing Knowledge

2.1.1 Introduction of RealCost

RealCost is a computer software which analyzes the life-cycle cost of highway projects. If various factors are entered into the software, LCCA is computed automatically and the software shows comparisons in regard to all alternatives of a project. This software has been designated as one analysis tool in FHWA for two major purposes: to provide an instructional and educational tool for decision makers in pavement design and to provide a practical tool for pavement designers who can use the results when developing pavement plans (FHWA 2004). The software compute life-cycle values for both agency and user costs at the same time with regard to rehabilitation projects as well as new construction. Also, it provides deterministic and probabilistic modeling of LCCA. One important function of the RealCost compares results of alternatives to analyze the advantages and disadvantages of many alternatives. However, even though this software provides analyzed cost data from various angles, the output from the RealCost isn't the only crucial factor for decision making since there are many other factors to consider such as risks, available budget, political situations and environmental issues. Therefore, even though RealCost provides critical information for the overall decision making process, as an economic analysis tool, the optimized output from RealCost is just one solution of many alternatives (FHWA 2004).

The California Department of Transportation (Caltrans) is one of the leading State Agencies (STAs) using and developing LCCA. RealCost manual issued by Caltrans provides background data which includes maintenance and rehabilitation cost during the pavement's life cycle by considering various regional, climatic and physical aspects of highways which are critical factors for computing life-time costs. In other words, the level of accuracy and classification of the background data determines the quality of LCCA.

2.1.2 Problems in Present Life-Cycle Cost Analysis

Poorly managed highway infrastructure is one of the public issues in the US even though highway infrastructure influences the national economy in many ways (Choi and Kwak 2011; Shatz et al. 2011). As one way to escape from the present recession, the Obama administration has decided to invest a huge amount of money to reform highway infrastructure and STAs are needed to increase efficiency in these budget allocation and management. As a decision support tool, LCCA has been used for comparing alternatives, decision making and budget allocation in many STAs (Salem et al. 2003). In addition, much research has been conducted to develop LCCA methodologies. However, STAs depend on background data for LCCA from their individual empirical data. There are no established criteria at the nationwide level, even though critical cost elements determining life cycle costs including routine maintenance costs are essential to provide more exact life time costs. Especially, routine maintenance costs have wide variability influenced by environmental,

regional and physical conditions of pavement in a number of ways. Thus, routine maintenance cost prediction methodologies should be developed based on various factors and should be studied to provide more exact prediction of life time cost for highway recovery projects.

2.1.3 Limitation of LCCA Tools

There have been many trials to establish systems to support decision making for new construction or rehabilitation highway projects such as LCCA tools, planning tools and design tools. Specifically, LCCA tools like RealCost provide useful economic information to decide highway design and choose among many alternatives. However, one problem is whether or not these LCCA tools have sufficient background data to support reliable results. The California Department of Transportation (Caltrans), leading LCCA development and usage, has developed M&R cost schedules but categories representing climate regions are divided into five sectors for maintenance and rehabilitation (M&R) schedule as shown in Table 1. Moreover, the planned maintenance costs are same year by year in the table of pavement M&R Schedules which is a major database for RealCost. As a result, these limitations make it difficult to plan annual budget allocations. Since LCCA is not only for decision making for initial projects, but for whole life cycle costs and cash flow analysis, this limitation has clearly become an obstacle for developing detailed and exact analyses for life cycle costs of a project.

Table 1. Categorized climate region used in California (Caltrans 2007)

California Climate Regions Categorized by Caltrans	Climate Regions for Pavement M&R Schedule Used in Caltrans
North Coast	All Coastal
Central Coast	
South Coast	
Inland Valley	Inland Valley
High Mountain	High Mountain & High Dessert
High Dessert	
Desert	Desert
Low Mountain	Low Mountain & South Mountain
South Mountain	

2.2 Research Objectives

Objective 1: Collect reliable highway maintenance data for Texas

Objective 2: Identify critical factors influencing pavement conditions

Objective 3: Identify relationships between routine pavement maintenance costs and
identify factors impacting pavement condition

Objective 4: Provide lookup tables for the routine pavement maintenance costs
categorized based on identified factors.

2.3 Research Methodologies and Hypothesis

2.3.1 Tasks to Achieve Research Objectives

Task 1: Identify physical and environmental factors impacting pavement condition.

Task 2: Collect reliable data for identified factors and maintenance cost

Task 3: Test the relationships between identified factors and maintenance cost

Task 4: Create a maintenance cost prediction model using statistical method

Task 5: Test validity of the model through comparison with other highway data

Task 6: Provide lookup tables based on each categorized factor.

2.3.2 Stochastic Analysis

To provide a reliable pavement maintenance prediction model, the statistical analysis method was applied in this research step by step. The data used in this study was from the *Pavement Management Information System* (PMIS) which has been widely used in TxDOT and its district agencies. The following statistical procedures were implemented with PMIS data.

1. Data classification based on the research standard

According to the research scope, data should be classified and manipulated. If many independent factors are identified and it's difficult to analyze the data trend under normal statistical methods, the data should be classified once more through clustering analysis: statistical classification methods.

2. Descriptive analysis and scatter plot

Descriptive analysis provides basic data features for extracted samples such as standard deviation, mean and variance. Also, scatter plots provide insight into the basic tendency of sample data.

3. Pearson and Spearman correlation test

This statistical test is to test the fundamental relationships between identified variables and the dependent variable.

4. Box-Cox analysis

If the relationship has high significance in the Spearman test, the sample data need transformation. Box-Cox analysis provides the best fit in the sample data transformation.

5. Casewise diagnostics and residual scatter plot

To increase the model's reliability, the outliers are eliminated through checking residual scatter plots and Casewise diagnostics. Also, equal variance for the sample data can be checked through standardized residual scatter plots.

6. ANOVA test

Once regression analysis is performed and produces prediction models, this test identifies the model's significant differences among means of variables.

7. Individual coefficient test and co-linearity analysis

The Coefficient test provides a method to check whether or not the individual dependent variable has significance with the suggested prediction model. Also, co-linearity analysis provides test methods if there are correlations between dependent variables.

2.3.3 Hypothesis

The specific purpose of this study was to identify:

1. If there is any relationship between identified physical and environmental factors and maintenance cost and
2. If it's possible to predict the maintenance cost or obtain statistical tendency between the identified factors and maintenance cost

Following model was used to test this hypothesis:

$$\text{Model: } Mc = \beta_0 + \beta_1 * At + \beta_2 * Ap + \beta_3 * Pt + \beta_4 * Tv + \beta_5 * Pa$$

At = Annual Average Temperature (°F)

Ap = Annual Average Precipitation (inch)

Pt = Pavement thickness (Surfacing Thickness) (ordinal value)

Tv = Traffic Volume based on Annual Average Daily Traffic (count)

Pa = Pavement Age from the last surface overlay date (years)

Mc = Annual Maintenance Cost per lane mile (\$/lane-mile)

The suggested variables tested those influences on pavement through the literature review and could be changed based on the process of statistical analysis. This model confirms relationship between annual pavement maintenance cost and critical factors which were identified in the literature review. Annual pavement maintenance cost would be predicted based on this equation after being developed and validated.

2.4 Research Assumptions

- Under \$10,000 of annual maintenance costs per lane mile in pavement thickness over 2.5 inches and under \$8,000 annual maintenance cost per lane mile in pavement thickness under 2.5 inches are regarded as routine maintenance costs without any overlay activities (Scullion et al. 1997).
- Regional climate factors such as precipitation and temperature are the same in one county (Scullion et al. 1997).
- Typical life expectancy of ACP is 10 years and within this life expectancy, material characteristics of ACP follow expected physical characteristics.
- The sample data is normally distributed according to the Central Limit Theorem. The population of data was from PMIS having over 190,000 section data and the sample data about 1,000 section data was extracted from PMIS.

2.5 Limitation

- The research is limited to State Highway (SH) sections in Texas
- The research is limited to only Asphalt Concrete Pavement (ACP)
- The research is limited to only routine maintenance costs, not including surfacing overlay or rehabilitation costs. For long pavement life, there are several recovery strategies which are classified differently by system, STAs and definitions. In this research, the routine maintenance concept was shown in Figure 1.

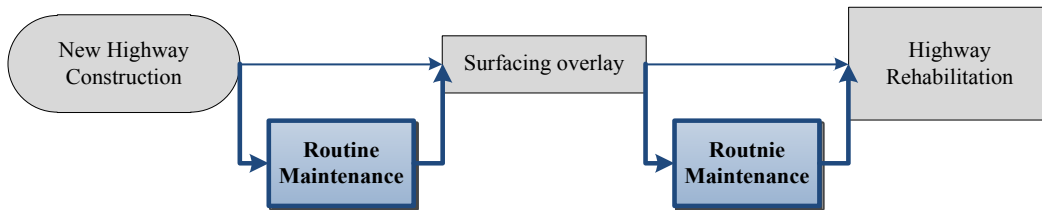


Figure 1. Highway maintenance work-frame

- The original source of data for this research is from TxDOT's Pavement Management Information System (PMIS) which is under development. Thus, imperfections in PMIS such as omissions in data have influenced the result of this study.

2.6 Significance of the Research

This research will be of significance to the highway agencies located in Texas since the objective is to provide detailed maintenance cost data based on identified factors which can be utilized in LCCA for their decision making and annual budget allocations. The research unveils whether or not raw factors such as environmental conditions and physical features of pavement affect maintenance costs representing pavement management practices. Moreover, the pavement maintenance cost prediction model is suggested by using statistical methodology which connects identified environmental factors and maintenance costs. That is, this study is one trial to make a connection between two fields of studies: engineering studies and management practices.

3 LITERATURE REVIEW

The purpose of the literature review was to understand related studies including engineering and maintenance practices for life-cycle-cost analysis of highway projects. Specifically, the literature review was conducted to identify critical factors impacting routine highway maintenance costs in Texas and to investigate current LCCA practices used for comparison and decision making between highway project alternatives.

3.1 Highway Facts

There are several types of roadway systems in the United States classified as functional system, ownership, location and federal-aid or no federal-aid roadways (FHWA 2008). Fundamental Highway classification is a functional classification system established in 1989 by the Federal Highway Administration (FHWA). This system comprises three blocks such as local roads, collectors and arterials. Sometimes, Freeway/Expressway is included in this categorization to separate heavy traffic highways from arterial roads. Local roads are for homes, businesses, farms and small communities and provide channels to collector roadways. The major function of collectors is to provide access from local roads to arterials and arterial roads connecting between towns and cities (FHWA 2011).

Among these highway systems, the FHWA has classified National Highway system (NHS) according to the importance of the national economy, defense and mobility.

All interstate highways (IH), the Strategic Highway Network, intermodal connectors and other principal arterials belong to NHS whose total length is about 166,000 miles but occupies only 4% of the US roadway systems and carries more than 40% of all highway traffic, 75% of heavy truck traffic and 90% of tourist traffic (Slater 1996).

Several government organizations have ownership of roadways: the Federal Agency, State Highway Agency (STA), County, Town or other jurisdictions. These institutions generally have rights and responsibilities of their owned roadways (FHWA 2008)

Roadway location is one of the standards to classify roadways into “Urban” area roadways and “Rural” area roadways. Since the increase and spread of population, there is a general trend to decrease rural area roadways (FHWA 2008).

Another way to categorize roadway is whether or not there is federal government support. Federal aid roadways include the National highway system; if the roadways don't have any federal supports, they are classified as non-federal-aid roadways. These classifications are combined to analyze many aspects and to make up annual statistical data for roadways in the United States.

In Texas, there are four major highway types such as Interstate Highway (IH), US route (US), State Highway/Route (SH or SR) and Farm to Market Road (FM).

Generally, IH and US are classified as National Highway System (NHS). SH and FM are classified as highways under local government agencies. Most highways including IH and US are owned by state governments (FHWA 2008), but management and design criteria for these highways is still under FHWA control. In case of SH and farm to market roads, various standards are under local government rules such as state, municipal and county governments. Particularly, these criteria for SH follow rules of the Texas state government and state highway agency.

Texas is one of the largest states in the United States. Accordingly, Texas has a huge highway system to manage which is total 79,000 lane mile. Paved lane-miles of major highways in Texas are as shown in Table 2.

Table 2. Total lane-mile of Texas highways in 2005 (Mikhail et al. 2006)

Highway type	Asphalt Concrete Pavement(ACP)	Continuously Reinforced Concrete Pavement (CRCP)	Jointed Concrete Pavement(JCP)	Total
IH	4,745	1,346	244	6,335
US	14,288	638	368	15,294
SH	15,566	819	434	16,819
FM	40,403	201	124	40,728
Total	75,002	3,004	1,170	79,176

3.2 Highway Management Studies

Many highways which have exceeded in their 20 years design life expectancy have created critical challenges for State Transportation Agencies. Even though most highways were constructed from 1960s to 1980s during a construction boom, much of the overage pavement in these highways is still in use (Lee and Ibbs 2005). In consequence, the improvement of transportation infrastructure became one of the fourteen *grand challenges for engineering in the 21st century* by the *National Academy of Engineering* under the Obama administration which passed the *American Recovery and Reinvestment Act of 2009* (ARRA act). The purpose of the ARRA act and selection of the 14 engineering challenges was to provide investment in infrastructure, education, health and green energy.

Out of these plans, the amount of federal investment in highway recovery projects is about \$80 billion but there are many serious problems related to the economy of new construction and traffic disruptions. So, with FHWA and State Highway agencies, the paradigm for highway construction has turned from new construction to maintenance and renewal of existing facilities represented as 4R: restoration, resurfacing, rehabilitation and reconstruction (Choi and Kwak 2011; Lee and Ibbs 2005).

In 2008, national total disbursements for physical maintenance of state administered highways were \$12,499,324,000. Out of this expenditure, Texas spent \$1,300,886,000 for roadway physical maintenance during one year (FHWA 2008). In

general, one district in Texas spends an average of \$52,000,000 per year which is the sum of combined contracted and non-contracted maintenance expenditures (TxDOT 2011). Clearly, each district invests huge amount of money annually to maintain highway system which is major transportation infrastructure.

Pavement maintenance is highly related to safety, serviceability and efficiency of highways. Pavement damage caused by traffic loads and environmental influences worsen the performance of highways. So, pavement maintenance on a regular basis is critical to maintain and improve pavement performance and quality (Embacher and Snyder 2001; Labi and Sinha 2005). Inadequate budget allocations for pavement maintenance regardless of heavy use and environmental conditions is often ineffective and cause wasted resources in spite of STAs' limited financial situation of agencies (Labi and Sinha 2005). Moreover, not only maintenance costs but rehabilitation costs increase due to failure of efficient decision making. Therefore, it's essential to maintain pavement in a timely and effective way (Chou and Le 2011).

Life Cycle Cost Analysis (LCCA) is one way to improve efficiency in highway construction activities including new construction as well as 4R projects (Labi and Sinha 2005). As an engineering economic analysis tool and a decision-support tool, LCCA is useful to compare project alternatives. From the Intermodal Surface Transportation Equity Act of 1991, implementing LCCA in highway investment decisions has been one of the main policies that FHWA encourages (FHWA 2002).

It's highly recommended that LCCA be completed as early as possible. Especially, in all California highway projects supervised by Caltrans, LCCA must be performed based on established procedures and data in the LCCA procedure manual issued by Caltrans. Fundamental steps for LCCA are as follows: 1. Establish Design Alternatives 2. Determine Activity timing 3. Estimate Costs including agency and user costs 4. Compute life-cycle costs 5. Analyze the results (FHWA 2004).

There are two different computational approaches in LCCA: Deterministic and Probabilistic. First, the deterministic approach is the recommended and traditional way for project decision-support (Caltrans 2007). In this approach, each input variables for LCCA should be fixed and separated. These variables can generally be determined based on historical evidence or professional judgment (Caltrans 2007). On the contrary, a probabilistic approach is a relatively new methodology based on uncertainty and variation with regard to input variables (FHWA 2002) . Compared with an individual computation of the deterministic approach, the probabilistic approach has advantages because different assumptions for uncertain variables based on probability distribution can be included in the computation at the same time. However, the probabilistic methodology has been under-developed until now and many STAs and FHWA don't recommend this method as yet (Caltrans 2007; FHWA 2002). So, Agencies allow only the deterministic approach for LCCA in highway projects at this time (Caltrans 2007).

To perform a LCCA in a certain project, necessary factors are as follows:

1. Design alternatives
2. Analysis period
3. Discount rate
4. Maintenance and rehabilitation sequences
5. Costs
6. RealCost software

Among these factors, costs include initial costs (construction costs and project support costs), maintenance costs, rehabilitation costs and user costs (Caltrans 2007). Also, maintenance costs are comprised of costs for routine, preventive and corrective maintenance such as sealing, void under sealing, chip sealing, patching, spall repair, individual slab replacement, thin hot mix asphalt overlay, etc (Stampley et al. 1993; TxDOT 2008). In case of Caltrans, annualized maintenance costs are used and this historical cost data is collected by the Division of Maintenance. Following Figure 2 is an example of maintenance costs used by Caltrans.

TABLE F-2 Inland Valley Climate Region HOT MIX ASPHALT PAVEMENT MAINTENANCE AND REHABILITATION SCHEDULE																										
Final Surface Type	Pmnt Design Life	Maint. Service Level	Year		Begin Alternative Construction		5	10	15	20	25	30	35	40	45	50	55									
New Construction/Reconstruction																										
HMA	20	1,2	Year of Action		0						18		23				33		38				48		53	
			Activity Description		New/ Reconstrct						CAPM HMA		Rehab HMA (10 yr)				CAPM HMA		Rehab HMA (10 yr)				CAPM HMA		Rehab HMA (10 yr)	
			Activity Service Life (years)	Annual Maint. Cost (\$/lane-mile) over Activity Service Life	18	3,600	5	1,100	10	3,200	5	1,100	10	5,200	5	\$1,100	10	5,200								
			Year of Action		0						18		23				41				46					
			Activity Description		New/ Reconstrct						CAPM HMA		Rehab HMA (20 yr)				CAPM HMA				Rehab HMA (20 yr)					
			Activity Service Life (years)	Annual Maint. Cost (\$/lane-mile) over Activity Service Life	18	3,600	5	1,100	18	2,700	5	1,100			18	2,700										
		3	Year of Action		0						18				27				36		43					
			Activity Description		New/ Reconstrct						CAPM HMA				CAPM HMA				CAPM HMA		Lane Replace (20 yr)					
			Activity Service Life (years)	Annual Maint. Cost (\$/lane-mile) over Activity Service Life	18	3,600	9	5,600	9	4900	7	5,700	18	3,600												

Figure 2. Example of typical pavement M&R schedule in Caltrans (Caltrans 2007)

The federal highway agency strongly recommends the use of LCCA in the early stage of highway construction projects and in some states LCCA is a mandatory process in the pre-construction stage of highway projects (FHWA 2002). Obviously, LCCA methodology provides a clear insight into life time cost allocations for highway agents who try to predict budget expenses and their allocation.

However, even the most advanced users of LCCA methodology have used a relatively undetailed maintenance costs index based on a few categorized climate regions (Table 1). Moreover, annual maintenance costs are the same all the time until and even after roadway rehabilitation. This means that these data ignore many factors such as traffic volume, pavement age and annual interest rate. With this data, it's incorrect to predict and allocate annual maintenance costs of each highway. Due to at least 20 years pavement life expectancy, annual maintenance cost is the main factor used to analyze more exact life time costs and annual budget allocations with both deterministic and probabilistic life-cycle cost analysis. Especially, to improve reliability for the probabilistic approach in highway LCCA, more detailed and categorized maintenance costs are necessary.

3.3 Maintenance Practice and Criteria

Texas Department of Transportation (TxDOT) has established several standards for pavement maintenance based on procedure, range, type and budget amount. These standards should be referred to whenever highway agencies make plans or decisions for highway maintenance. First of all, pavement maintenance should be implemented based on the following six phases: planning, budgeting, scheduling, performing, reporting and evaluating (DYE Management Group 2006). All maintenance processes should be carried out according to highway maintenance plans established and revised on an annual basis to obtain more realistic plans.

Fundamentally, there are three kinds of plans categorized by length of term: annual plans, four year plans and long-range transportation plans. Annual plans primarily focus on actual pavement management. In contrast to the annual plan, the four year plan and the Texas long-range transportation plan are conceptual and directional plans for improving all transportation systems over designated periods (Gao et al. 2011; Scullion et al. 1997; TxDOT 2009; TxDOT 2010).

TxDOT categorizes maintenance work into three areas: routine maintenance, preventive maintenance and major maintenance. Most preventive and major maintenance work should be contracted but routine maintenance works can be performed by either the agency's force or contract (Embacher and Snyder 2001; TxDOT 2008). In addition to this, each maintenance scope is managed by a different

division or program with a different budget range and different work scope as shown on the following table.

Table 3. Maintenance categories defined in Maintenance Management Manual
(TxDOT 2008)

	Routine Maintenance	Preventive Maintenance	Major Maintenance
Contract System	RMC* should be developed through CMCS**	Should be programmed in advance as CPM***	Should be programmed in advance as MMP****
Division	District Agency	Transportation Planning and Programming Div.	Transportation Planning and Programming Div.
Level	District	State	State
Project cost	Less than \$300,000 /PJ	No limitation	No limitation
Work Scope (Travel way)	Pavement repair Crack seal Seal coat Level-ups Light overlay(≤ 2") Additional base	Milling Crack seal Seal coat Level-ups Light overlay (≤ 2") Microsurfacing	Stabilize base Subgrade, add base Seal coat Level-ups Light overlay (≤ 2") Widening

RMC*: Routine Maintenance Contract

CMCS**: Construction/Maintenance Contract System

CPM***: Contract Preventive Maintenance

MMP****: Major Maintenance Program

Within PMIS, four maintenance categories are defined: preventive maintenance, light rehabilitation, medium rehabilitation and heavy rehabilitation or reconstruction (Stampley et al. 1993). This category is different from the three categorized maintenance work scopes defined in the *Maintenance Management Manual* issued

by TxDOT in 2008. In addition, the work scope and cost of each maintenance category follow the rules of PMIS itself.

Table 4. PMIS maintenance types and costs defined in PMIS (Stampley et al. 1993)

Maintenance Type	Last Overlay type classified by thickness		
	4 (Hot-Mix Asphalt Overlay over 5.5")	5 (Hot-Mix Asphalt Overlay 2.5"-5.5")	6 (Hot-Mix Asphalt Overlay below 2.5")
Preventive Maintenance (PM)	Crack Seal or Surface Seal \$10,000 / lane-mile	Crack Seal or Surface Seal \$10,000 / lane-mile	Crack Seal or Surface Seal \$8,000 / lane mile
Light Rehabilitation (LRhb)	Thin Asphalt Overlay \$35,000 / lane-mile	Thin Asphalt Overlay \$35,000 / lane-mile	Thin Asphalt Overlay \$35,000 / lane-mile
Medium Rehabilitation (MRhb)	Thick Asphalt Overlay \$75,000 / lane-mile	Thick Asphalt Overlay \$75,000 / lane-mile	Mill & Asphalt Overlay \$60,000 / lane-mile
Heavy Rehabilitation (HRhb)	Remove Asphalt Surface Replace & Rework Base \$180,000 / lane-mile	Remove Asphalt Surface Replace & Rework Base \$180,000 / lane-mile	Reconstruct \$125,000 / lane-mile

Like the above mentioned systems, maintenance scope and range differ according to the purpose of the maintenance systems. Particularly, the concept and work scope of preventive maintenance defined in PMIS are similar to the routine maintenance work scope in the *Maintenance Management Manual* except for light overlay, level-ups and additional base work. These different scopes and standards can possibly cause inefficiency and confusion in an agency's work processes. Inversely, the large number of systems established to develop efficiency in managing highways creates many possibilities for inefficiency, inversely. Thus, clear and united standards seem

to be necessary to increase system unity and decrease the complexity inherent in connecting between huge numbers of systems.

Since the data source for this research came from PMIS and the objectives of this research -routine maintenance cost prediction except overlay- were closer to the concept of PMIS, the data scope for their research analysis has followed the PMIS standard.

3.4 Highway Engineering Studies

As described in Figure 3, fundamentally, there are three basic components in the pavement structure: foundation, road-base and surfacing. In a flexible-type pavement, the surfacing includes wearing course and base course and the foundation includes sub-base and subgrade (Rogers 2003).

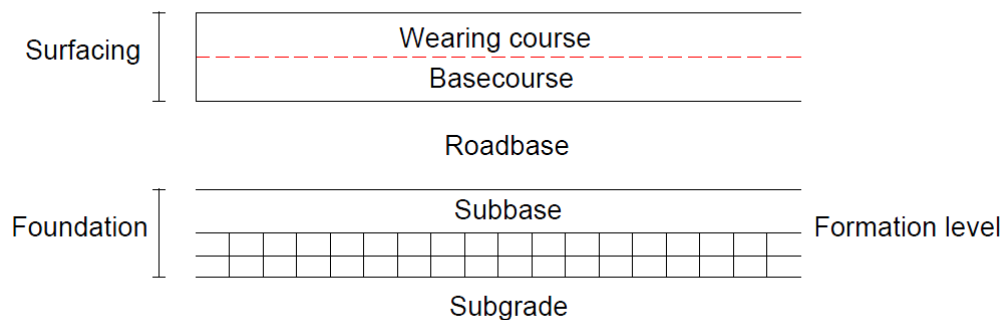


Figure 3. Flexible pavement structure (Rogers 2003)

Of these layers, there are two types of surfacing: flexible type and rigid type. Normally, flexible type means asphalt based pavement (ACP) and rigid type means concrete based pavement (CRCP). However, in PMIS, pavement type was classified in three ways: ACP, CRCP and Jointed Concrete Pavement (JCP). Basically, CRCP and JCP were included in continuous reinforced concrete pavement (CRCP) (TxDOT 2011)

Since these pavement structures are constantly exposed to daily traffic and outside environment, when agents design a highway project they must consider various factors as follows: pavement performances, traffic, roadbed soil, materials, environments, drainage, reliability, life-cycle costs and shoulder design (AASHTO 1993; Birgisson et al. 2000). Much research has been conducted to investigate the relationship between these factors and pavement conditions.

To display roadway conditions in a quantitative way, there is a representative pavement condition index, Pavement Condition Index or Rate called PCI or PCR which is often used in pavement management to measure pavement distresses (Yu et al. 2008). Aside from PCI, there are many methods and indices for measuring according to the purpose of measurement. For example, the *Pavement Serviceability Index* (PSI) is one way to show pavement serviceability including distress and roughness (AASHTO 1993) and the *International Roughness Index* (IRI) determines ride quality parameters by measuring pavement smoothness (TxDOT 2011). In

addition, there are different kinds of indices such as *Remaining Service Life* (RSL) which predicts the expected life of the pavement until its condition doesn't provide normal highway functions any more (Yu et al. 2008). Many agents, institution and researchers choose the indices suitable for their purpose and have been recording the result of evaluations based on the indices in their system to obtain and manage information of each roadway condition. Also, considerable engineering research related to pavement has been used these indices.

Based on PCI and RSL in asphalt concrete pavement (ACP), four parameters have significant relationships with pavement performances (Jeong 2008) and survival time of pavement: temperature, precipitation, overlay thickness, traffic loading (Yu et al. 2008). The following shows the research results of the four parameters vs. pavement survival life. This study assumes that PCI 70 is the end of the serviceability of the pavement.

3.4.1 Survival Life Versus Overlay Hot Mixed Asphalt (HMA) Thickness

The increase of overlay HMA in every 1 inch (2.5 Cm) results in about a half year extension of pavement service life. Moreover, for thicker HMA overlays (more than 2 inches), the increase of pavement thickness extends service life. This means that thicker pavement surfacing can be a favorable choice if the financial situation is favorable (Yu et al. 2008).

As described in Figure 4, asphalt thickness is significantly related to pavement temperature which is one of the most important factors in pavement damage. Normally, thick pavement records slightly lower temperatures than thin pavement, resulting in strong resistance to climatic factors (Wahhab and Balghunaim 1994).

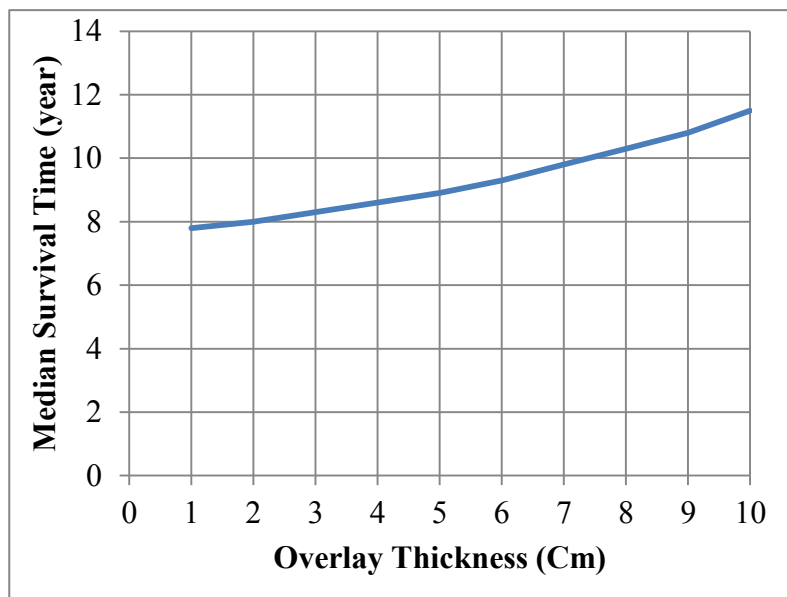


Figure 4. Survival time versus Overlay thickness (Yu et al. 2008)

3.4.2 Pavement Performances Versus Temperature

In design ACP, pavement temperature is a mandatory input variable to predict pavement performance which was developed by the *Strategic Highway Research Program* (SHRP). Pavement temperature is primarily influenced by air temperature (Solaimanian et al. 1993).

In many cases, an increase in pavement temperature causes pavement performance deterioration through loss of strength in ACP. In some cases, climatic loading becomes a far more influential factor than physical loading such as traffic loading (Salter and Al-Shakarchi 1989; Solaimanian et al. 1993; Wahhab and Balghunaim 1994)

However, in low temperature, ranges from 47 °F to 55 °F, a 1°C rise (1.8°F) in annual average temperature causes a one year pavement service life extension. However, these data from Ohio, whose annual average temperature is from 47 °F to 55 °F, might not be suitable for areas having temperature ranges over 60 °F (Yu et al. 2008).

Although there are many research has been based on different temperature ranges and, accordingly, provided different research results, it's clear that temperature is a critical influence on pavement service life (Solaimanian et al. 1993; Yu et al. 2008).

3.4.3 Pavement Performances Versus Precipitation

Another climate factor, precipitation, also plays an important role in pavement performance (Saraf et al. 1987) and survival time. A four inches (10 Cm) increase in precipitation means a decrease as much as one year in service life. So, it can be said that an increase in annual rainfall worsen pavement conditions. Since these two climatic factors, precipitation and temperature, are the characteristics of a region which agents should consider, highway agents can apply these factors for budget allocations for highways (Rogers 2003; Yu et al. 2008).

Particularly, two climatic factors, temperature and precipitation, impact pavement performance in various ways. The pavement deterioration procedure is as follows (Kapiri et al. 2000):

1. Temperature gradients result in significant physical stress in the slabs
2. Temperature differences in pavement create cracks or joint openings
3. Moisture variation in pavement causes differential shrink from top to bottom

That is, inherent pavement temperature, which is closely affected by air temperature, increases pavement cracks and moisture originated from regional rainfall or humidity catalyzes this deterioration process (Kapiri et al. 2000). Climatic factors have the most influence on pavement condition compared to physical load (Kapiri et al. 2000).

3.4.4 Survival life Versus Traffic loading

In the pavement design guide in FHWA, AASHTO and STAs, traffic loading has been the major factor to be considered in new highway construction. Relationship of these two factors is displayed in Figure 5. Actually, a ten-times increase in average daily traffic leads to a half year decrease in pavement service life. That is, degeneration in pavement conditions from heavy traffic may be canceled to some degree by increased pavement thickness (Yu et al. 2008).

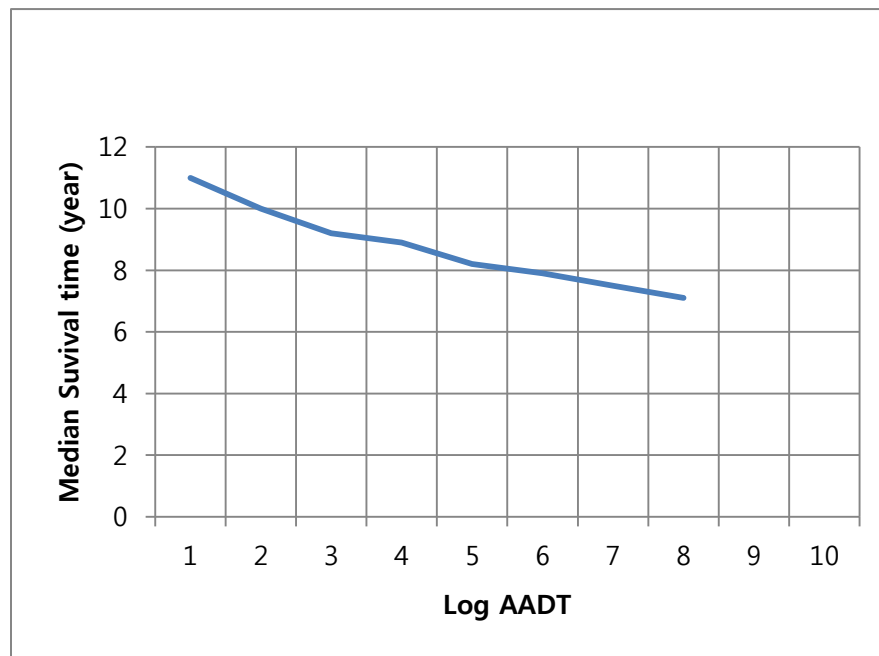


Figure 5. Survival time versus Log-AADT (Yu et al. 2008)

The PMIS manual issued by TxDOT designated several factors to identify to determine if pavement maintenance is necessary: 1. Pavement type. 2. Distress ratings. 3. Ride Score. 4. Average daily traffic per lane (ADT). 5. Roadway

functional class. 6. Average county rainfall (in inches per year). 7. Time since last surface (in years) (Stampley et al. 1993). These factors were defined as critical factors in PMIS which must be included when pavement treatment is considered.

These research results show that environmental and physical factors have critical impacts on pavement condition. Of course, these studies obviously have limitations to generalize to all highways in the US because the results reflect only the characteristics of highways in specific states and regions. This means that according to the nature of other states, there are possibilities for different research results. However, pavement physical features are clearly influenced by these results and moreover, future research has been introduced in this research to derive direct cost predictions from these factors (Yu et al. 2008).

4 DATA COLLECTION AND STRATIFICATION

4.1 Introduction of Pavement Management Information System (PMIS)

In the early 1980's the *Texas Department of Transportation* established the *Pavement Evaluation System* known as PES. Around 1990, a *Pavement Management Steering Committee* was organized to plan improvements to PES. They established two main objectives for future PES systems: First, the system should show network level pavement conditions, impact analysis and fund allocation which make it possible to document pavement condition and identify maintenance and rehabilitation candidates at a District level. Second, the system should be integrated into all levels of decision making within TxDOT. The first objective was achieved when PES was upgraded to PMIS but the second objective is now being studied and implemented (Scullion et al. 1997).

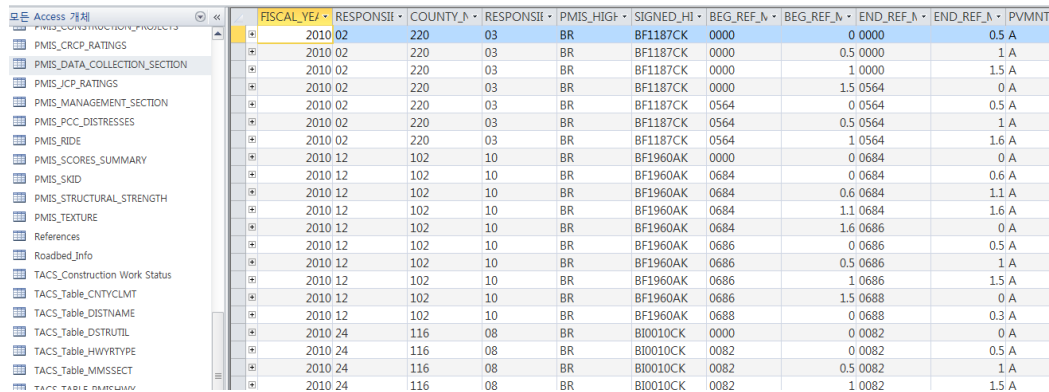
PMIS includes and categorizes various factors since highway managers need to check actual pavement conditions, past investment amounts, maintenance action history, section information, pavement physical data, traffic loading, etc. The factors which should be identified to treat pavement designated in the PMIS are as follows: First, managers should identify pavement type among three broad pavement types such as ACP, CRCP and JCP, and ten classified pavement types numbered from 1 to 10 (Stampley et al. 1993). Then, identify pavement conditions with evaluation standards and equations provided in PMIS such as pavement score and deterioration curves. This is displayed as pavement distress ratings and ride scores. Third, check

the traffic loading of pavement with annual average daily traffic per lane and highway functional class. Fourth, check average county rainfall and finally, last surface date should be checked representing pavement age (Scullion et al. 1997).

Even though continuous and systematical trials have existed to make PMIS perfect, there are still many problems identified at the district level. Inversely, agencies' requests for the system's improvement proves that PMIS is a critical resource in their pavement management efforts. These problems in the system can be classified into four groups: data collection, data analysis score, output format and training. Especially, problems in data collection are critical issues which can possibly threaten reliability and usefulness of PMIS. Data scoring problems also can make the data impractical due to inefficient and inconsistent criteria (Scullion et al. 1997). These potential problems originating from PMIS can cause a limitation in this research such as a limitation of available samples. However, clearly, PMIS is the most reliable system out of many other established systems and is now improving. In addition, most district level and county level agencies have produced and utilized PMIS data as a main source for decision making for pavement maintenance projects.

4.2 Data Stratification

PMIS collates all highway section information based on designated parameters, this data can be exported to a Microsoft Access file as shown in Figure 6.



FISCAL_YEAR	RESPONSIBILITY	COUNTY_ID	RESPONSIBILITY	PMIS_HIGHWAY	SIGNED_HIGHWAY	BEG_REF_ID	BEG_REF_ID	END_REF_ID	END_REF_ID	PVMNT
2010 02	220	03	BR	BF1187CK	0000	0 0000	0 0000	0 0000	0 0000	0.5 A
2010 02	220	03	BR	BF1187CK	0000	0.5 0000	0.5 0000	0.5 0000	0.5 0000	1 A
2010 02	220	03	BR	BF1187CK	0000	1 0000	1 0000	1 0000	1 0000	1.5 A
2010 02	220	03	BR	BF1187CK	0000	1.5 0564	1.5 0564	1.5 0564	1.5 0564	0 A
2010 02	220	03	BR	BF1187CK	0564	0 0564	0 0564	0 0564	0 0564	0.5 A
2010 02	220	03	BR	BF1187CK	0564	0.5 0564	0.5 0564	0.5 0564	0.5 0564	1 A
2010 02	220	03	BR	BF1187CK	0564	1 0564	1 0564	1 0564	1 0564	1.6 A
2010 12	102	10	BR	BF1960AK	0000	0 0684	0 0684	0 0684	0 0684	0 A
2010 12	102	10	BR	BF1960AK	0684	0 0684	0 0684	0 0684	0 0684	0.6 A
2010 12	102	10	BR	BF1960AK	0684	0.6 0684	0.6 0684	0.6 0684	0.6 0684	1.1 A
2010 12	102	10	BR	BF1960AK	0684	1.1 0684	1.1 0684	1.1 0684	1.1 0684	1.6 A
2010 12	102	10	BR	BF1960AK	0684	1.6 0686	1.6 0686	1.6 0686	1.6 0686	0 A
2010 12	102	10	BR	BF1960AK	0686	0 0686	0 0686	0 0686	0 0686	0.5 A
2010 12	102	10	BR	BF1960AK	0686	0.5 0686	0.5 0686	0.5 0686	0.5 0686	1 A
2010 12	102	10	BR	BF1960AK	0686	1 0686	1 0686	1 0686	1 0686	1.5 A
2010 12	102	10	BR	BF1960AK	0686	1.5 0688	1.5 0688	1.5 0688	1.5 0688	0 A
2010 12	102	10	BR	BF1960AK	0688	0 0688	0 0688	0 0688	0 0688	0.3 A
2010 24	116	08	BR	BI0010CK	0000	0 0082	0 0082	0 0082	0 0082	0 A
2010 24	116	08	BR	BI0010CK	0082	0 0082	0 0082	0 0082	0 0082	0.5 A
2010 24	116	08	BR	BI0010CK	0082	0.5 0082	0.5 0082	0.5 0082	0.5 0082	1 A
2010 24	116	08	BR	BI0010CK	0082	1 0082	1 0082	1 0082	1 0082	1.5 A

Figure 6. PMIS data exported to Microsoft Access

This study utilized 2010 PMIS data, the latest version. Fundamentally, PMIS data includes almost all available data, including 103 factors, in terms of highway identifiers, pavement physical status factors, influence factors and management data. Thus, these data should be sorted based on identified factors having critical impact on pavement.

4.2.1 Identified Factors

Based on engineering literature reviews, four of the most critical factors influencing pavement physical status were identified: pavement thickness, annual average temperature, annual average precipitation and traffic loading. In addition to these factors, last surfacing date should be checked before a maintenance project is

planned (Stampley et al. 1993). According to the PMIS manual issued by TxDOT, seven factors should be considered before proposing treatment: Pavement type, distress ratings, ride score, ADT per lane, functional class, average county rainfall and time since last surface (Stampley et al. 1993). In this research, time since last surface was expressed as “pavement age” and counted as a year unit. Thus, precipitation was considered as a county average value according to PMIS criteria. In addition, annual average temperature, another climatic factor identified in the literature review, was also considered as a county average value in this research. In the case of ride score or distress score, which is the engineered result for pavement condition, it was not considered since one objective of this research was to investigate impacts from environmental and physical factors on pavement maintenance cost. Therefore, identified factors in this research were fivefold: traffic loading represented as AADT, annual average precipitation, annual average temperature, pavement thickness and pavement age. Required data for this research can be extracted from the PMIS database.

4.2.2 Data Stratification Procedure

Basically, PMIS includes a broad range of data from management data to physical information of all highway sections in Texas. The data pool categorizes over 100 items and includes over information for 39,000 sections of state highways (SH). This wide range of data should be manipulated based on research objectives.

In the first Phase, necessary items relating to identified critical factors were classified as follows:

Step 1. Sort out items including basic highway identifier showing locations, sections and highway systems.

Step 2. Based on the research objectives and identified factors, items should be sorted out. The classified items are displayed on Table 5.

Table 5. Data sorted from PMIS (DYE Management Group 2006)

Items	Definition	Example, Format or unit
PMIS_HIGHWAY_SYSTEM	Broad category of highways used in PMIS to simplify analysis and reporting.	US/IH/SH
SIGNED_HIGHWAY_RDBD_ID	This field includes the highway system, highway number, highway suffix, and the roadbed ID.	SH0087 K
PVMNT_TYPE_BROAD_CODE	Identifies the broad category of pavement.	A, C, J
PVMNT_TYPE_DTL_RD_LIFE_CODE	Code indicating predominate travel lane pavement type during the data collection year of the data collection section.	01~10, 99
SECT_LNGTH_RDBD_OLD_MEAS	Roadbed mileage for the data collection section. This field will be the same as section-length-centerline initially.	Typically 0.5
AADT_CURRENT	The published average daily estimate of vehicles for all lanes of traffic on a particular highway	
FUNCTIONAL_SYSTEM	A general description of the type of service that the PMIS data collection section is intended to provide over time	01~19
MAINTENANCE_COST_AMT	The cost of pavement maintenance done on the main travel lanes during the previous year of data collection for the data collection section.	\$
NUMBER_THRU_LANES	Total number of thru-lanes for a section of highway	
LAST_OVERLAY_DATE	Date of the last overlay on the data collection section.	yyyymm

The Second phase for stratification classifies the data according to the research range involving research objectives, limitations and assumptions established in advance.

The data range classification procedure was implemented as follows:

Step 1 . This research investigated only in the state highway (SH) system.

Step 2 . Only asphalt concrete pavement (ACP) type was included in the analysis.

Step 3 . Main road systems were considered. Among roadbed identification numbers in PMIS, the main-lane roadway system is expressed as one character, K, L, R.

Step 4 . Among many functional classifications, this research focused only on the arterial roadway system.

The Third phase was converting data from the classified data (which had raw values) into appropriate range unit and research standards. This manipulation process was based on the equations displayed below:

1. $\text{Section lane-mile} = \text{Section Length} \times \text{Number of Lanes}$

2. $\text{Maintenance Cost per lane mile}$

$$= \text{Maintenance Cost Amount of Section} \div \text{Section lane mile}$$

3. $\text{Pavement Age (year): total year from last overlay date to December 2010}$

$$= (2010 - \text{last overlay year}) + (12 - \text{last overlay month})/12$$

4. $\text{Pavement Thickness Value Converting}$

Table 6. Pavement thickness converting criteria

Actual Pavement Thickness	Pavement Thickness Values in PMIS	Converted Value (Ordinal Value)
$T > 5.5''$	04_Thick ACP	3
$2.5'' < T \leq 5.5''$	05_Medium ACP	2
$T \leq 2.5''$	06_Thin ACP	1

Adjusting the manipulated data range to provide actual research data was the fourth phase of data stratification. Based on the research assumptions and limitations, the classified data were redefined and arranged. The process according to the criteria was as follows:

Step 1. Based on assumptions, life expectancy of asphalt surfacing pavement was 10 years (120 months). So data over pavement age over 120 months were excluded.

Step 2. Based on an objective of this research to focus only on routine maintenance. According to the literature review on maintenance practices, the data were selected from the following criteria.

- a. Maintenance cost range in thin asphalt (< 2.5 inches) was \$ 8,000 per lane-mile.
- b. Maintenance cost range in medium and thick asphalt surfacing was \$ 10,000 per lane-mile.

Actually, after completing the fourth phase, the stochastic analysis was begun to create a full range prediction model. However, to improve model reliability, a more

advanced stratification method was developed. Thus, through a statistical grouping method, “*Clustering Analysis*,” the manipulated data was divided into each computed group and then a regression analysis for each group was implemented. The five phase procedures for data stratification is displayed in Figure 7.

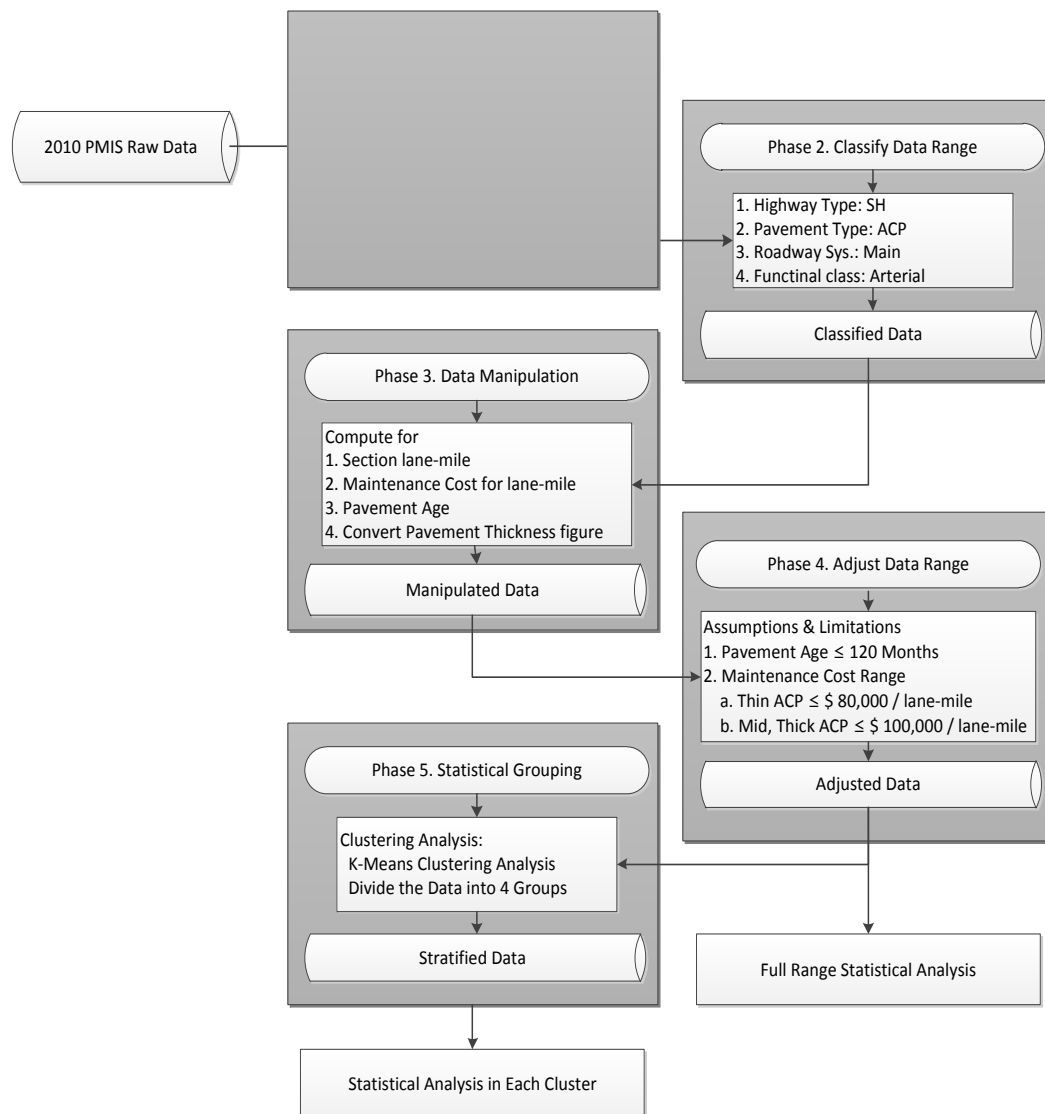


Figure 7. Data stratification procedure

Table 7. Data range after phase 2 classification

	Data Range	
Number of Sections	954 sections	
Number of Counties	33 counties	
Number of Highway system	34 different system	
Temperature Range	8.6 °F	63°F – 71.6 °F
Precipitation Range	28 inches	29 inch – 57 inch
Pavement age range	25.67 year	0.83 year – 26.5 year

5 STOCHASTIC MODELING FOR FUTURE PAVEMENT MAINTENANCE COST

5.1 Regression Analysis with Full Range Data

5.1.1 Descriptive Statistics

Table 8 shows the descriptive statistics results of the adjusted data set. The dependent variable in this research was maintenance cost amount per lane mile (\$/lane-mile) marked by Mc. The mean value of Maintenance cost was \$ 988.6 per lane-mile with standard deviation, of \$1886.41 per lane mile. Annual average temperature (At) had a mean value of 66.4 °F with 2.39 °F standard deviation. Annual average precipitation represented as Ap was 36.4 inches (approximately 920.5 mm) with a standard deviation 7.88 inches. Pavement thickness (Pt) value was converted to ordinal values. The real thickness value is as follows: 1 (Thin, less than 2.5 inches), 2 (medium, from 2.5 inches to 5.5 inches), 3 (Thick, not less than 5.5 inches). PMIS provides thickness data in this way so converting was required for statistical analysis. Annual average daily traffic (AADT) had a mean value of 4516.35 (number of vehicles per day) and a standard deviation of 3500.99. The mean value of pavement age was 85.99 months with a standard deviation of 27.67 months.

Table 8. Descriptive Statistics

	Variables	Unit	Mark	Mean	Std. Deviation	N
Dependent var.	Cost	\$/lane-mile	Mc	988.6	1886.41	487
Independent var.	Temperature	°F	At	66.4	2.39	487
	Precipitation	Inch	Ap	36.4	7.88	487
	Thickness		Pt	1.75	0.693	487
	AADT	No. of vehicle/day	Tv	4516.35	3500.99	487
	Age	years	Pa	85.99	27.67	487

5.1.2 Scatter Plots

The Scatter plot of maintenance cost (Mc) vs. annual average temperature (At) in Figure 8 shows a positive relationship. That is, temperature increase affects the increase in pavement maintenance costs. The slope is 145.58, meaning that if the temperature increases by 1 °F, the actual increase in maintenance cost will be \$145.58 / lane-mile per year. Moreover, P-value of this individual regression is 0.001, which also indicates the significance of this relationship.

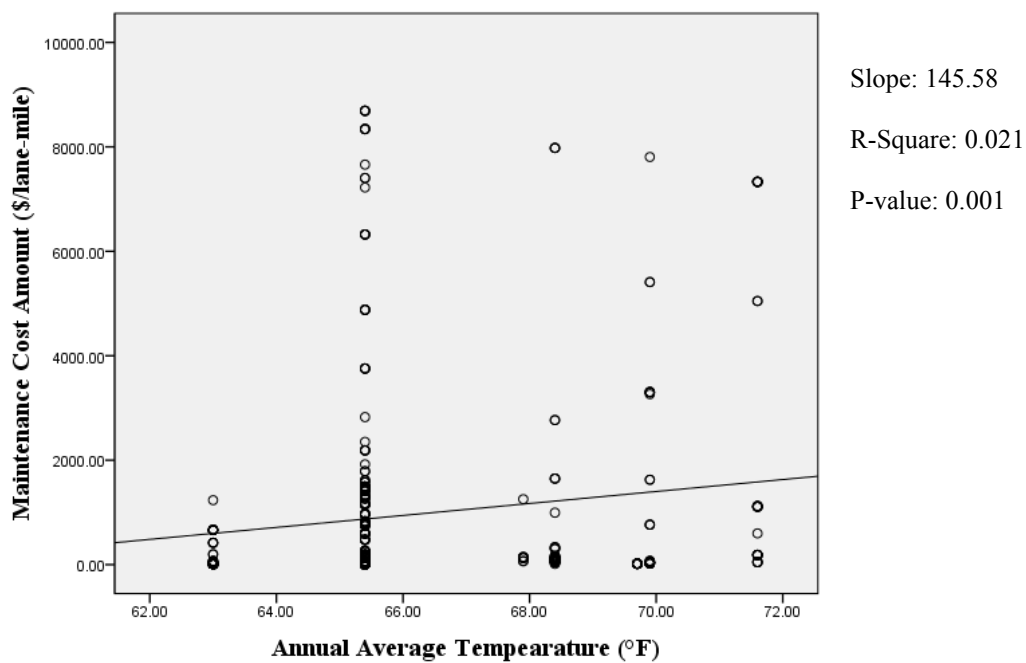


Figure 8. Scatter plot of maintenance cost versus annual average temperature

The scatter plot Figure 9 shows maintenance cost (Mc) vs. annual average precipitation (Ap). This plot shows a negative relationship between two variables and the slope naturally has a negative value, -14.007, which means, one inch increase in annual average precipitation decreases maintenance cost as much as \$14 / lane-mile. P-value is 0.198 which represents lower significance between the two variables.

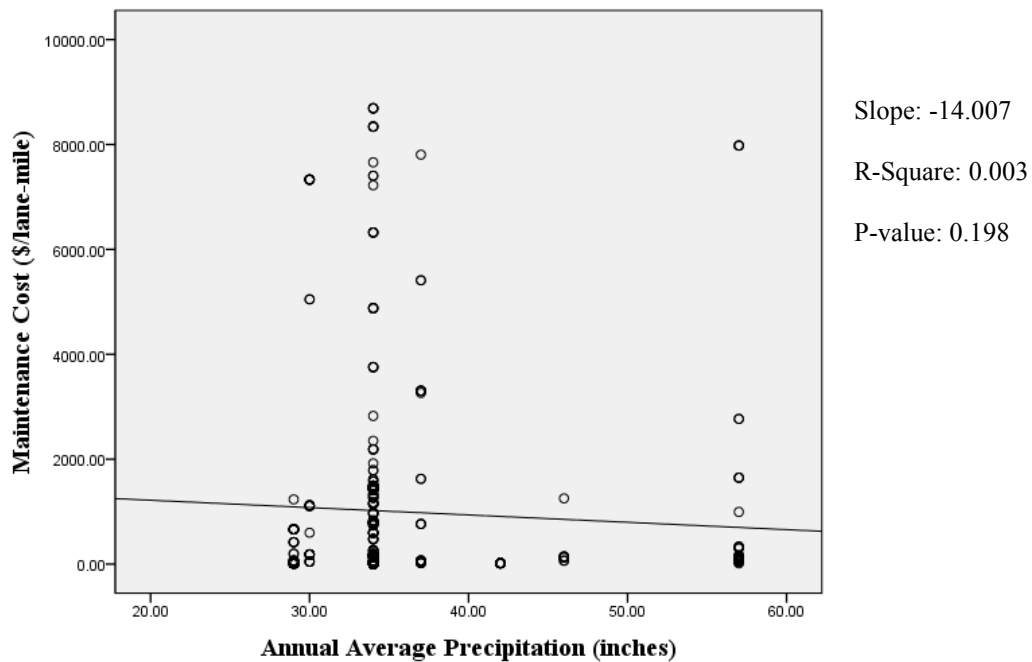


Figure 9. Scatter plot between maintenance cost versus annual average precipitation

The scatter plot in Figure 10 shows maintenance cost (Mc) vs. pavement thickness (Pt). As previously mentioned in the data stratification chapter, pavement thickness is not numerical scale, but an ordinal scale. So, in this scatter plot, the relationship can be checked according to the ordinal scale of pavement thickness. The general tendency of this scatter plot shows that if pavement thickness is increased, the maintenance cost will increase as well, at a rate of \$ 223.006 / lane-mile. The p-value between them is 0.000 based on Spearman's test, representing significance in the relationship. However, this trend is beyond normal expectations, so more advanced data analysis is required.

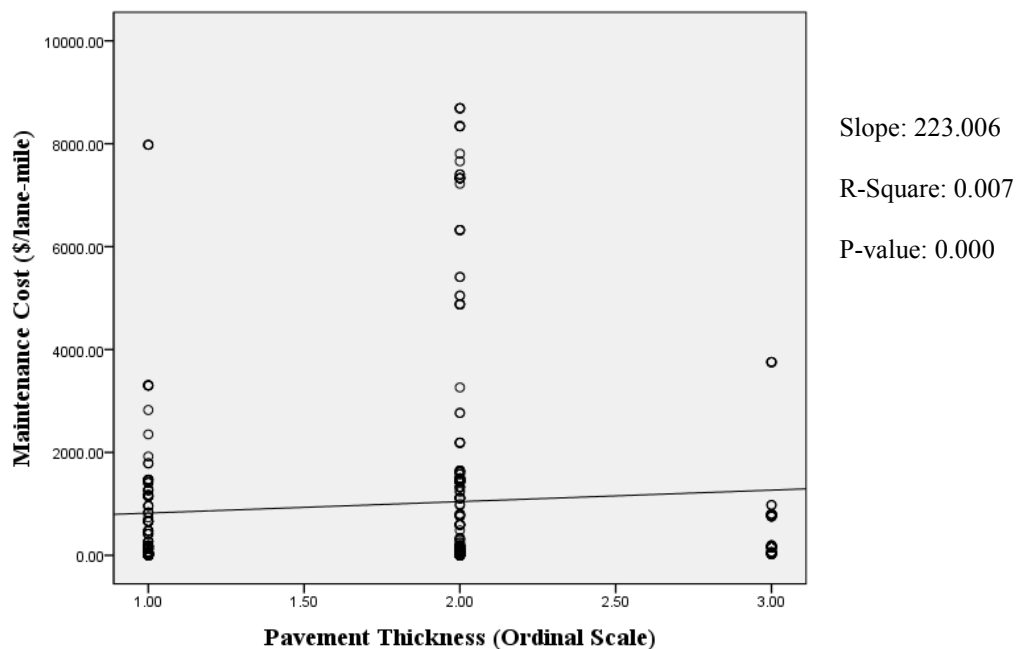


Figure 10. Scatter plot between maintenance cost versus pavement thickness

Figure 11 represents the relationship between maintenance cost (Mc) and Traffic volume (Tv). This output was also beyond common sense since this analysis output displayed a negative relationship between them as much as -0.017, meaning that if AADT increases, maintenance cost was decreased. Also, the p-value of this relationship is 0.49, representing almost no relationship. This data also required considerable re-classification due to low significance.

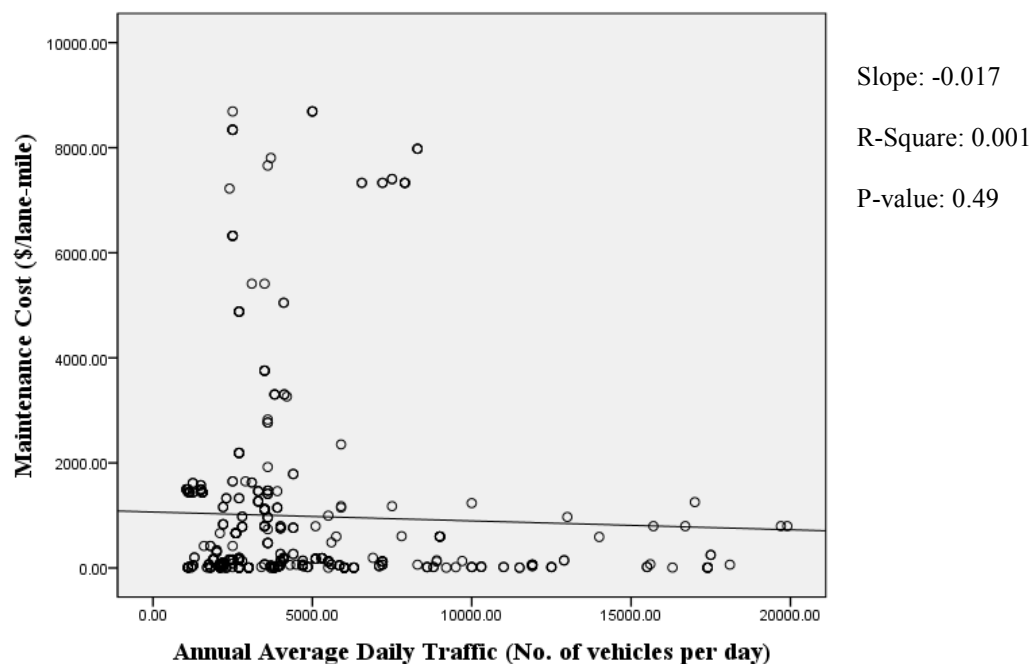


Figure 11. Scatter plot between cost versus traffic volume

The relationship between maintenance cost (Mc) vs. pavement age (Pa) is displayed in Figure 12. This indicates that there is a positive relationship between them by an increase of \$2.217 /lane-mile per month. P-value of 0.474 also indicates almost no relationship between them

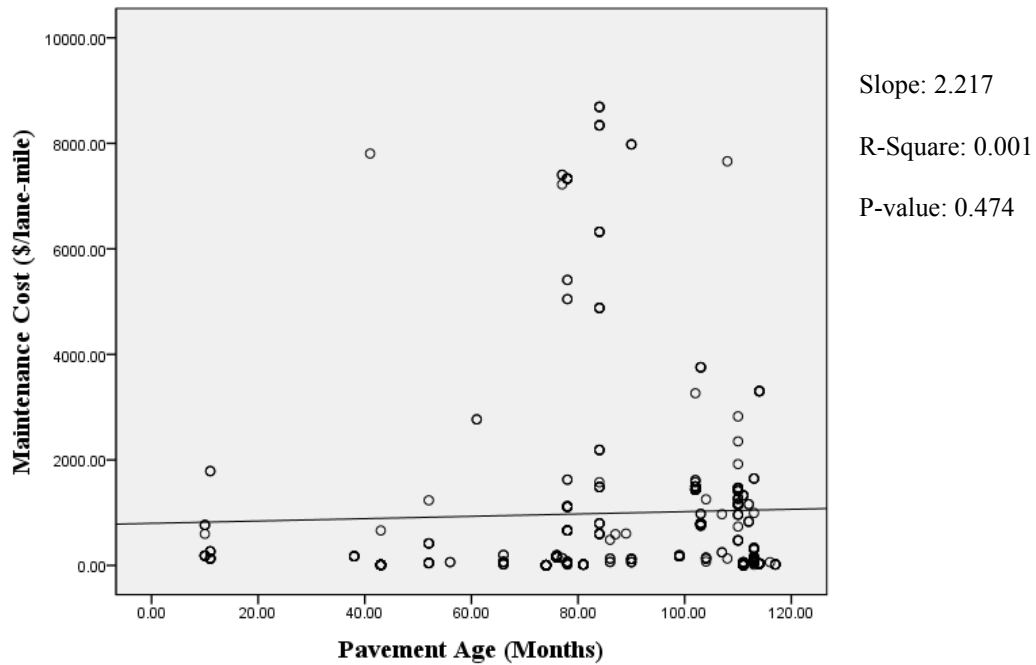


Figure 12. Scatter plot between maintenance cost versus pavement age

5.1.3 Correlation Analysis

Table 9. Correlation analysis

			Mc	Tv	At	Ap	Pt	Pa
Pearson	Mc	Correlation	1	-.031	.145**	-.058	.082	.033
		Sig. (2-tailed)		.490	.001	.198	.071	.474
		N	487	487	487	487	487	487
Spearman	Mc	Correlation	1	-.105*	.128**	.010	.193**	.061
		Sig. (2-tailed)		.021	.005	.827	.000	.175
		N	487	487	487	487	487	487

** . Correlation is significant at the 0.01 level (2-tailed).

* . Correlation is significant at the 0.05 level (2-tailed).

Table 9 shows the overall correlation between maintenance cost and each variable. Since thickness is the only ordinal scale value, this correlation test uses the Spearman test. In this individual correlation test, temperature and pavement thickness have significant relationships with maintenance cost. Also, each correlation coefficient represents whether the relationship follows a negative or positive trend through checking the coefficients' signs.

5.1.4 Multiple Regression Analysis

Since correlation analysis and R square values indicate that there is no significant relationship, before performing the regression analysis, outliers at the level of ± 1 standard deviation residual were excluded like the residual plot in Figure 13. This standardized residual level is a bit robust to produce a reliable prediction model.

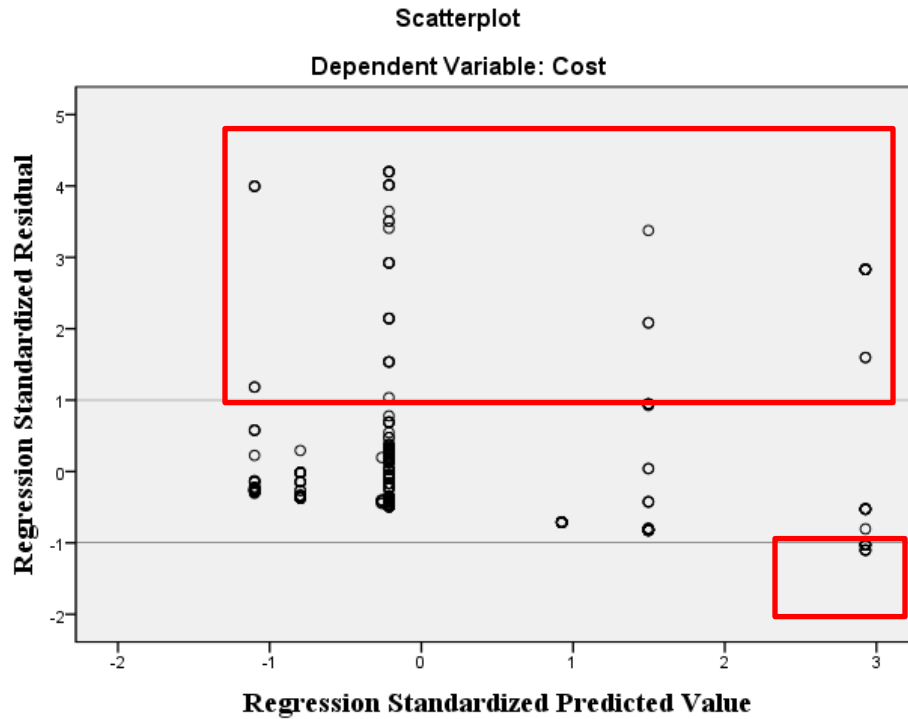


Figure 13. Residual plot for outlier detection

These outliers were detected by performing “Casewise Diagnostics” which show all standard deviations of residuals. As a result, 54 samples were excluded from the data set.

Table 10. Regression Model Summary

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate	Change Statistics				
					R Square Change	F Change	df1	df2	Sig. F Change
1	.373 ^a	.139	.129	607.80214	.139	13.836	5	427	.000

As displayed in Table 10, the adjusted R square of the model is 0.129, which means that this model can explain as much as 12.9% the data. Above all, with this small adjusted R-square value, it is difficult to produce a reliable maintenance cost prediction model. In this regression model, the residual plot is fan-shaped as shown in Figure 14, meaning that the data need transformation.

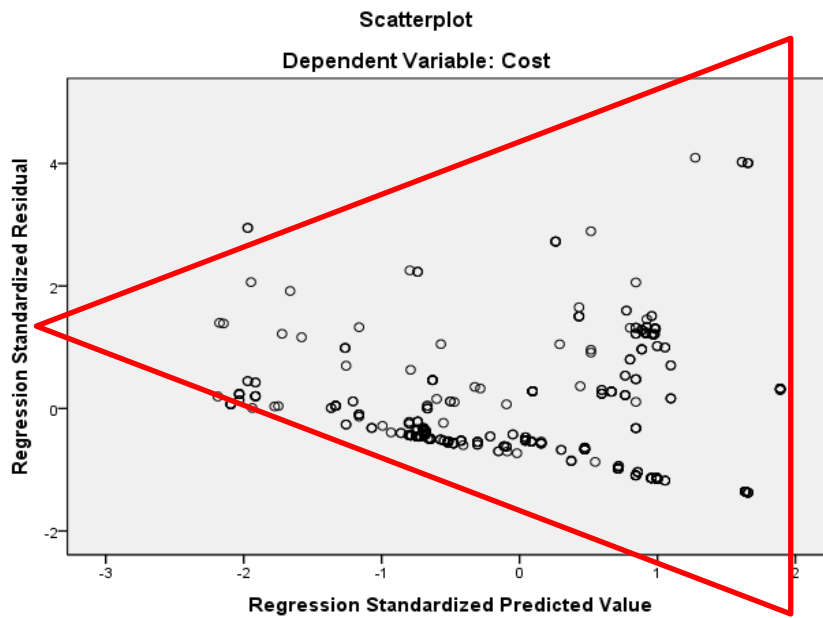


Figure 14. Residual plot of initial regression analysis

Box-Cox analysis was employed to gain a reliable transformed value. Figure 15 shows the result of Box-Cox analysis. Accordingly, the transformation value is as follows:

$$\text{Transformed Maintenance cost} = (\text{Maintenance cost})^{-0.1}$$

lambda	rmse	ci95
- .10	349.13	344.69
.10	351.05	344.69
-.20	368.35	344.69
.20	373.44	344.69
-.30	401.89	344.69
.30	412.90	344.69
-.40	451.73	344.69

Figure 15. Transformed results analyzed from Box-Cox

Table 11. Transformed regression summary

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate	Change Statistics				
					R Square Change	F Change	df1	df2	Sig. F Change
1	.296 ^a	.087	.078	.12657	.087	9.224	5	481	.000

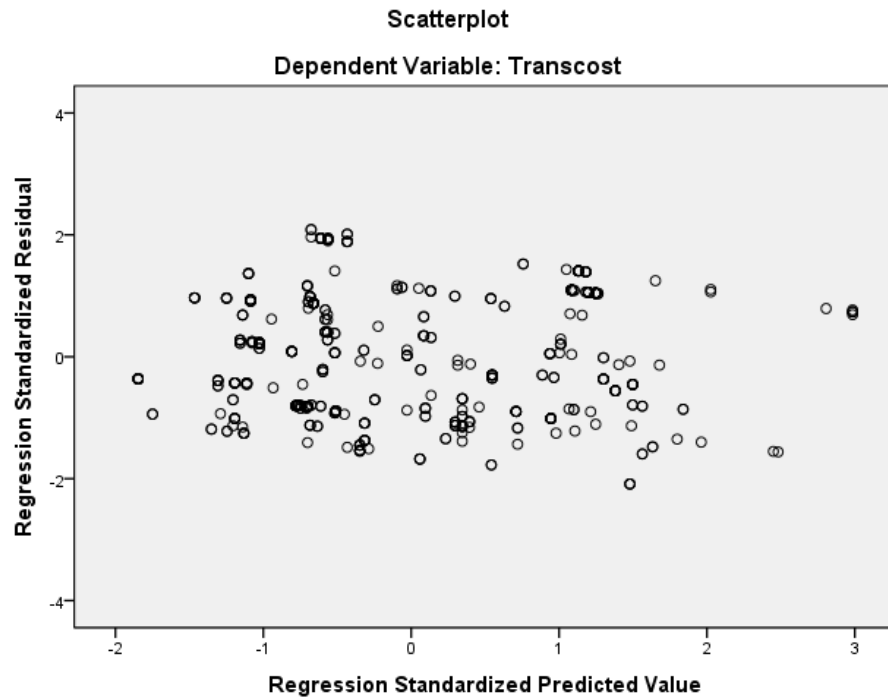


Figure 16. Transformed regression residual plot

After transforming the dependent variable, the fan-shaped residual distribution was removed and created an equal distribution. This change can be checked through residual scatter plot in Figure 16. However, the R-square value decreased to 0.078. That is, it is almost impossible to explain the data set with this model.

5.1.5 Discussion

With just stratified data, the full-range regression model proved that it is not appropriate to predict costs with sufficient reliability. These difficulties could be explained in two ways: 1) it can be assumed that there are other unexpected elements impacting pavement maintenance costs such as natural disasters and vehicle

accidents. These unexpected elements are very likely to cause outliers in the data set which reduce reliability of the model and R-square value. However, since the outliers were eliminated at standard residual level ± 1 before performing the regression analysis, this assumption is unrealistic. 2) At certain points, the relationship between dependent and independent variables could changed dramatically. To predict maintenance cost with reliability, the “inflection points” or “stable intervals” should be observed and also, each regression model in each stable interval or between inflection points should be established, respectively. To analyze these intervals, this research utilized the clustering analysis method.

5.2 Clustering Analysis

The goal of clustering analysis is to identify homogenous subgroups of cases in samples. In other words, cluster analysis can identify groups minimizing group variation and maximize between group variations (Aldenderfer and Blashfield 1984). The first step of this analysis is to identify similarities using the Euclidian distance between samples. This analysis method defines group similarity with the distance between samples and dissimilarity also with the distance between groups. Especially, in the *K-means cluster method*, the number of groups can be designated before beginning analysis according to the researcher’s assumptions or convenience in managing data. Therefore, the clustering method was employed in this research. The result of the K-means clustering analysis is as follows:

Table 12. K-means clustered results

Cluster	Tv (AADT) range	No. of Samples
Cluster 1	$1000 \leq \text{Cluster 1} \leq 3500$	239
Cluster 2	$3500 < \text{Cluster 2} \leq 7200$	179
Cluster 3	$7200 < \text{Cluster 3} \leq 13,000$	51
Cluster 4	$13,000 < \text{Cluster 4} < 20,000$	18

At first, 4 groups according to the AADT interval of 5,000 were expected. However, even though the grouping result was followed the sequence of Tv, the Tv range is not exact 5,000 interval. In addition, clusters 3, 4 had small number of samples which is hard to expect reliable analysis since data ranges in other factors are similar. Possibly, these problems would be originated from limited number of samples. Thus, in this research, AADT range for analysis was decided from 1,000 to 7,200.

The fact that the clustering was arranged based on Tv, shows that Tv has a pivotal impact on the regression model, in contrast to the low significance of Tv in full-range regression model. In other words, since 3 inflexion points originated from Tv make the data trend change dramatically, it can be assumed that R-square value in full-range regression model was displayed as almost no relationship.

5.3 Regression Analysis in Each Cluster

5.3.1 Regression Analysis for Cluster 1

As shown in Table 13, the correlation values and their significance is much higher in Spearman's rho test than the Pearson test. Pearson test is only for linear relationships between dependent and independent variables so it's hard to say that there is some linear relationship between maintenance cost (Mc) and five parameters. However, in the Spearman test, overall correlation values and their significance show a close relationship.

Table 13. Correlation analysis results for cluster 1

			At	Ap	Pt	Tv	Pa
Pearson	Mc	Correlation	.105	-.079	.138*	.162*	.115
		P-value	.105	.223	.033	.012	.076
		N	239	239	239	239	239
Spearman	Mc	Correlation	.312**	.267**	.292**	.118	.190**
		P-value	.000	.000	.000	.069	.003
		N	239	239	239	239	239

** . Correlation is significant at the 0.01 level (2-tailed).

* . Correlation is significant at the 0.05 level (2-tailed).

Based on the Spearman correlation analysis result, overall parameters have a significant relationship with Mc except Tv. The p-value in Tv is slightly lower than 0.05 significance level but, according to the engineering literature review identifying traffic volume's significant role in pavement conditions, Tv parameter was included

in this analysis. Also, overall correlation values of parameters have a positive relationship with the dependent variable. This is a positive tendency where if the parameters value increases, the dependent variable also increases.

In addition, high correlations in the Spearman test indicate that designated variables are non-linearly related. Generally, the Spearman test is conducted when the data value is based on “ordinal value” and moreover, the Spearman correlation method can be tested when it is hard to expect a linear relationship. That is, high significance in the Spearman test represents a non-linear relationship. So, the data need to be transformed to find the best fit model.

	lambda	rmse	ci95	var
1	.10	380.95	384.02	
2	.20	393.62	384.02	
3	-.10	404.56	384.02	
4	.30	424.10	384.02	
5	-.20	441.49	384.02	
6	.40	475.03	384.02	
7	-.30	497.63	384.02	
8	.50	550.64	384.02	
9	.10	576.64	384.02	

Figure 17. Transformed results analyzed by Box-Cox

Therefore, the dependent variable, Mc was transformed according to the Spearman's correlation test result by using the Box-Cox analysis tool shown in Figure 17. As a result, the transformation was implemented as follow:

$$\text{Transformed Mc} = (\text{Mc})^{0.1}$$

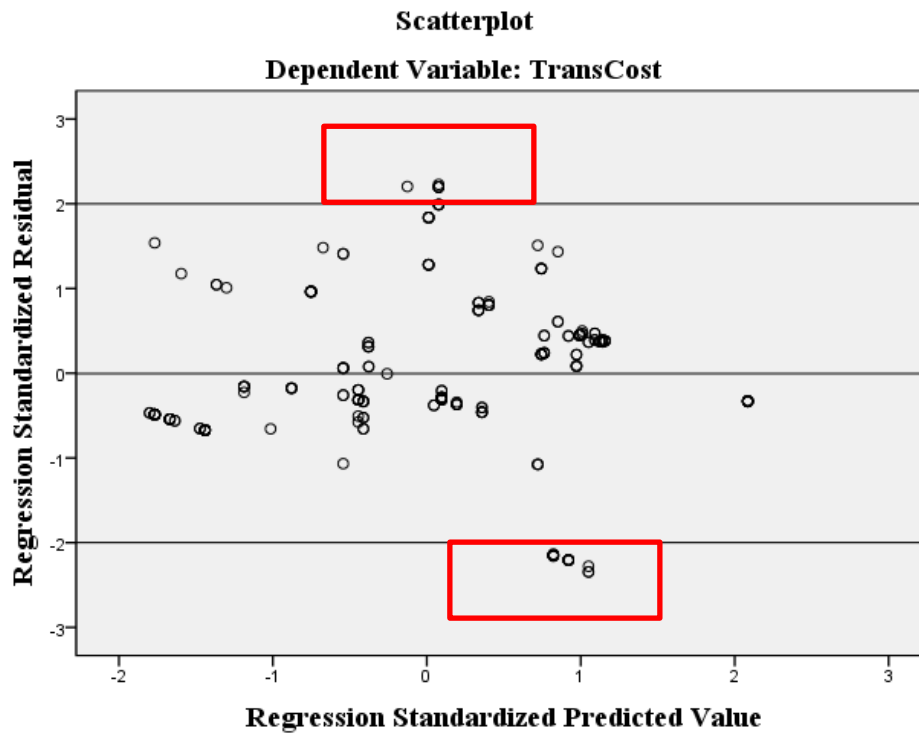


Figure 18. Residual plot of cluster 1

The residual plot shown in Figure 18 displayed random distribution, meaning that the model's equal distribution assumption can be accepted. Also, randomly distributed scatter represents that there is no specific tendency in the data set. Before conducting an analysis, to improve the model's reliability, the outlier displayed in the red box was eliminated in advance. Casewise diagnostics was employed to detect the outlier at a level of ± 2 . Table 14 shows a partial example of this process.

Table 14. Outlier detection process through Casewise Diagnostics

Casewise Diagnostics ^a				
Case Number	Std. Residual	Trans Mc	Predicted Value	Residual
100	1.538	1.91	1.4165	.49812
118	-2.346	1.17	1.9341	-.75944
119	-2.346	1.17	1.9341	-.75944
120	-2.279	1.20	1.9341	-.73783
128	2.204	2.43	1.7179	.71353

a. Dependent Variable: Transformed Mc

Table 15. ANOVA Table of regression model for cluster 1

Model		Sum of Square	Df	Mean Square	F-value	P-value	Adjusted R-square
a	Regression	15.499	5	3.100	83.996	.000 ^a	0.662
	Residual	7.639	207	.037			
	Total	23.138	212				
b	Regression	15.498	4	3.874	105.487	.000 ^b	0.663
	Residual	7.640	208	.037			
	Total	23.138	212				
c	Regression	15.482	3	5.161	140.881	.000 ^c	0.664
	Residual	7.656	209	.037			
	Total	23.138	212				

a. Predictors: (Constant), Pa, Tv, Pt, Ap, At

b. Predictors: (Constant), Pa, Pt, Ap, At

c. Predictors: (Constant), Pa, Ap, At

d. Dependent Variable: Mc

Table 16. Coefficients of transformed model for cluster 1

Model	Variable	Beta	Standard Error	t-value	p-value	VIF
a	Constant	-.665	.526	-1.263	.208	
	At	.036	.009	3.968	.000	2.488
	Ap	-.029	.002	-13.197	.000	2.396
	Pt	.014	.022	.634	.527	1.404
	Tv	3.082E-6	.000	.141	.888	1.448
	Pa	.154	.009	16.220	.000	2.197
b	Constant	-.693	.484	-1.433	.153	
	At	.037	.008	4.504	.000	2.002
	Ap	-.029	.002	-13.936	.000	2.173
	Pt	.014	.022	.663	.508	1.374
	Pa	.154	.009	16.340	.000	2.179
c	Constant	-.769	.470	-1.636	.103	
	At	.038	.008	4.873	.000	1.855
	Ap	-.029	.002	-14.278	.000	2.109
	Pa	.156	.009	17.510	.000	1.952

The backward elimination method was employed in this research to identify the appropriate independent variables. In cluster 1, c-model was selected since p-values of coefficient in Pt and Tv indicate no-significance, meaning that these variables have hardly any relationship to the transformed dependent variable in the suggested

regression model. This unexpected result will be discussed in the interpretation chapter.

As shown in the above ANOVA table and coefficient table, c-model was selected and its independent variables included just three parameters: Pa, Ap, and At. For this reason, a constant coefficient represented as β_0 was not considered in this model at the same time because the p-value is out of the significance level. By rejecting the null hypothesis - the regression model has no relationship between variables- the p-value in the ANOVA table was 0.000 representing that there is a significant linear relationship between the transformed dependent and independent variables. Moreover, the adjusted R-square value is 0.664, meaning that 66.4% of the variability in the transformed Mc can be explained by the 3 parameters: Pa, Ap and At. The remaining 33.8% of the variability might be explained by unknown factors. However, this study focused only on the relationship between Mc and the five independent variables. Finally, there is no multi-collinearity in this c-model for cluster 1; The Variance Inflation Factor (VIF) ranges from 1.885 to 2.109.

Based on the c-model, Equation (1) was established as a basic model equation to predict pavement maintenance cost with designated parameters. The selected model can explain 66.4% variability of the transformed maintenance cost (Mc). Accordingly, Equation (2) was a modified form of Equation (1) to predict maintenance cost with original unit, dollar-per-lane mile.

$$(M_c)^{0.1} = 0.038 * A_t - 0.029 * A_p + 0.156 * P_a \quad (\text{Equation 1})$$

$$M_c = \{0.038 * A_t - 0.029 * A_p + 0.156 * P_a\}^{10} \quad (\text{Equation 2})$$

The meanings of coefficients in the model are as follows:

1. β_1 is 0.038 meaning that if the annual average temperature (A_t) increases by one degree Fahrenheit, maintenance cost (M_c) will increase as much as \$0.038 per lane-mile as a transformed value.
2. β_2 is -0.29 meaning that if the annual total precipitation (A_t) increases by one inch, maintenance cost (M_c) will decrease as much as \$0.029 per lane-mile as a transformed value.
3. β_5 is 0.156 meaning that if pavement age (P_a) increases by one year from last pavement surfacing overlay time, maintenance cost (M_c) will increase as much as \$0.156 per lane-mile as a transformed value.

5.3.2 Regression Analysis for Cluster 2

Table 17. Correlation analysis results for cluster 2

			At	Ap	Pt	Tv	Pa
Pearson	Mc	Correlation	.251**	.277**	.163*	.316**	.225**
		P-value	.001	.000	.030	.000	.002
		N	179	179	179	179	179
Spearman	Mc	Correlation	.295**	.258**	.125	.197**	.283**
		P-value	.000	.000	.095	.008	.000
		N	179	179	179	179	179

** . Correlation is significant at the 0.01 level (2-tailed).

* . Correlation is significant at the 0.05 level (2-tailed).

Table 17 shows the correlation values and the significance of the Pearson's test and Spearman's test results which have similar correlation values and p-values. However, the p-value and correlation value of Tv is much higher in the Pearson's test which is one of the most important variables impacting pavement conditions. So, in this cluster, transformation was not used.

According to the Pearson's correlation analysis result, all parameters have a significant relationship with Mc. However, the correlation of Pt is checked by the Spearman's test because it's an ordinal value. The significance level in Pt is a bit higher than 0.05, meaning that Pt should be rejected but, according to the

engineering literature review showing Pt's major impact on pavement condition, the Pt parameter was included in this analysis.

All correlation values of independent variables have a positive relationship with the dependent variable. This means that if the parameter value increases, the dependent variable (Mc) will increase.

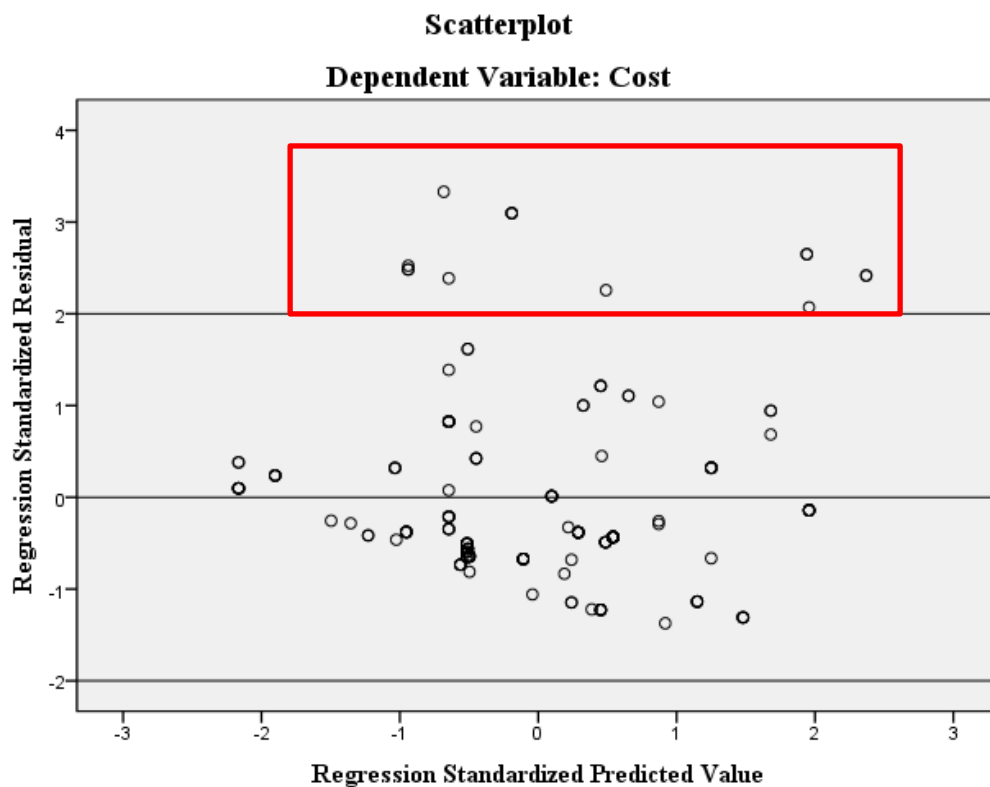


Figure 19. Residual plot of cluster 2

The residual plot shown in Figure 19 displayed a random distribution, meaning that the model's equal variance assumption can be accepted. Also, randomly distributed scatter represents that there is no specific tendency in the data set. To improve

model's reliability, the outlier displayed in the red box was eliminated in advance. Casewise diagnostics was also employed to detect the outliers at a level of ± 2 . Table 19 shows this process.

Table 18. Outlier detection process through Casewise Diagnostics

Casewise Diagnostics^a				
Case Number	Std. Residual	Mc	Predicted Value	Residual
16	2.650	4331.00	1935.0379	2395.96212
17	2.650	4331.00	1935.0379	2395.96212
18	2.417	4331.00	2145.2009	2185.79914
19	2.417	4331.00	2145.2009	2185.79914
61	3.097	3690.00	889.7907	2800.20930
62	3.097	3690.00	889.7907	2800.20930
63	3.097	3690.00	889.7907	2800.20930
84	3.330	3660.00	649.0009	3010.99908
100	2.387	2825.00	666.6567	2158.34334
131	2.525	2806.00	523.2137	2282.78626
133	2.257	3264.00	1223.3778	2040.62223
164	2.483	2768.00	522.9893	2245.01075
165	2.483	2768.00	522.9893	2245.01075
166	2.070	3815.00	1942.9771	1872.02289

a. Dependent Variable: Mc (Maintenance Cost)

Table 19. ANOVA table of regression model for cluster 2

Model		Sum of Square	Df	Mean Square	F-value	P-value	Adjusted R-square
a	Regression	2.750E7	5	5499029.083	17.490	.000 ^a	0.335
	Residual	4.999E7	159	314406.381			
	Total	7.749E7	164				
b	Regression	2.749E7	4	6871613.206	21.989	.000 ^b	0.339
	Residual	5.000E7	160	312495.670			
	Total	7.749E7	164				

a. Predictors: (Constant), Pa, Tv, Pt, Ap, At

b. Predictors: (Constant), Pa, Tv, Ap, At

c. Dependent Variable: Mc

Table 20. Coefficient of Transformed model for cluster 2

Model	Variable	Beta	Standard Error	t-value	p-value	VIF
a	Constant	-4405.054	1293.475	-3.406	.001	
	At	41.586	19.435	2.140	.034	1.060
	Ap	21.906	7.981	2.745	.007	1.391
	Pt	10.413	62.622	.166	.868	1.063
	Tv	.197	.049	4.040	.000	1.295
	Pa	7.547	1.376	5.485	.000	1.254
b	Constant	-4404.958	1289.539	-3.416	.001	
	At	41.762	19.347	2.159	.032	1.057
	Ap	21.698	7.859	2.761	.006	1.357
	Tv	.198	.048	4.164	.000	1.246
	Pa	7.592	1.345	5.645	.000	1.206

The backward elimination method was also employed in this analysis for choosing appropriate independent variables. In cluster 2, two models were suggested but the individual coefficient p-value in model a. was much higher than the 0.05 significance level. So, model a. was rejected and a second model was selected because the parameters in model b. had a p-value less than 0.05. In cluster 2, which has a higher AADT level than cluster 1, Tv was additionally included in the model but Pt, pavement thickness, was still rejected in the traffic range from 3500 to 7200.

As shown in the above ANOVA table and coefficient table, the model included four independent variables: At, Ap, Tv, and Pa including constant variable. The p-value of less than 0.05 in the ANOVA table represents that there is a significant linear relationship between the transformed dependent and independent variables by rejecting the null hypothesis: the regression coefficient has no relationship to the dependent variables. The adjusted R-square value is 0.339, meaning that 33.9% of variability in the transformed Mc can be explained by all the parameters: At, Ap, Tv, and Pa. The remaining 76.1% of the variability might be explained by unknown factors. However, this study focused only on the relationship between Mc and the five independent variables. Finally, there is no multi-collinearity in this c-model for cluster 1; The Variance Inflation Factor (VIF) ranges from 1.057 to 1.357.

Based on the model, Equation (3) was established as a basic model equation for cluster 2 to predict pavement maintenance cost with the selected four parameters.

The selected model can explain 33.9% variability of the transformed maintenance cost (Mc) with the original unit, dollar-per-lane mile.

$$Mc = -4404.96 + 41.762 * A_t + 21.698 * A_p + 0.198 * T_v + 7.592 * P_a$$

(Equation 3)

The meanings of coefficients in the model are as follows:

1. β_1 is 41.762 meaning that if annual average temperature (A_t) increases by one degree Fahrenheit, maintenance cost (Mc) will increase as much as \$41.762 per lane-mile.
2. β_2 is 21.698 meaning that if annual total precipitation (A_p) increases by one inch, maintenance cost (Mc) will increase as much as \$21.698 per lane-mile
3. β_4 is 0.198 meaning that if traffic volume (T_v), representing AADT, increases by one vehicle, maintenance cost (Mc) will increase as much as \$0.198 per lane-mile.
4. β_5 is 7.592 meaning that if pavement age (P_a) increases by one year from the last overlay date, maintenance cost (Mc) will increase as much as \$7.592 per lane-mile.

5.3.3 Discussion

1. Regression model for cluster 1.
 - Traffic volume of the cluster 1 model ranged from 1,000 to 3,500
 - The regression model includes three parameters: At, Ap, and Pa.
 - The regression model excludes traffic volume (Tv) and pavement thickness (Pt) parameters with low significance level of the coefficients.
 - With three parameters, the adjusted R-square value of the model is 0.664.
 - Overall meaning of the model is that if the values of At and Pa increase, the maintenance cost will increase and if Ap value increases, the maintenance cost will decrease.
 - The relationship between Ap and Mc varied from the expected output. In the literature review, Pa normally catalyzes the damage of the pavement condition but this research results showed a different trend. These results will be discussed in the interpretation chapter.

2. Regression model for cluster 2.
 - Traffic volume of the cluster 1 model ranged from 3500 to 7200.
 - Regression model includes all parameters: At, Ap, Tv and Pa
 - With all five parameters, adjusted R-square value of the model is 0.339
 - With the increased traffic volume level, one variable, Tv was included in the model for cluster 2 and the R-square values with four parameters decrease.

- Overall, meaning of the model shows that if A_t , A_p , T_v and P_a values increase, actual maintenance cost will also increase.

6 MODEL'S APPLICABILITY VALIDATION

One of the objectives in this research was to provide lookup tables for easy prediction of annual pavement maintenance costs. To achieve this goal, this study implemented a statistical analysis to confirm the correlation between variables and to produce a reliable prediction model by using regression analysis. However, the adjusted R-squares in the two clusters were quite different. Cluster 1 showed quite a relevant relationship but in cluster 2, it is hard to explain the maintenance cost with a somewhat low R-square value even though this value shows the variables having some relationship. This is the reason to produce lookup tables only for cluster 1. Therefore, a model validation test was implemented only for cluster 1 as well.

The PRESS (Predicted Error Sum of Square) statistic is one of statistical analysis methods applied in regression model for the purpose of testing suggested model's validity. Generally, obtaining and sorting out new data for validation of a proposed model is time-consuming and impractical. The PRESS statistic resolves these situations by providing statistical methodologies for the regression model validation (Ott and Longnecker 2008).

The PRESS statistic compares each observed response to the value based on the fitted model. This process can be expressed as follow:

$$\text{PRESS} = \sum_{i=1}^n (y_i - \hat{y}_i^*)^2$$

Originally, the PRESS statistic cannot be less than the value of SSE (Sum of Squared Error), but if the value of the PRESS statistic is close to the value of SSE, it proves that the proposed model can predict new data with high feasibility. In other words, large value of the PRESS statistic compared to SSE shows a validation problem in the proposed regression model.

In this research, Table 21 shows the values of the PRESS statistic and the SSE.

Table 21. The PRESS statistic versus the SSE of the cluster 1 model

	PRESS statistic	SSE
Model for Cluster 1	0.9763	0.9458

The result of the PRESS statistic has close value of SSE which proves that the model for cluster 1 can effectively predict new data under same data range. That is, not only the proposed model for cluster 1, but also lookup tables based on this model obtained the model's applicability validation.

7 CONCLUSION

This research focused on routine maintenance practices through analyzing cost factors with five external and physical parameters whose influence on the physical condition of pavement have already been investigated in many studies. Based on these studies, this research intended to discover whether or not environmental and physical features of pavement impact maintenance practices. PMIS was selected to analyze these relationships since PMIS is a reliable system managed by TxDOT which has tried to establish an information system for pavement maintenance and to enhance the system by supporting many academic projects to improve PMIS.

As has been suggested in previous engineering research, analyzing how management practices are influenced could provide valuable information and background for maintenance strategies. In addition, if the research finds significant relationships, this can provide a quantitative example about the relationship between engineering and management practices.

7.1 Interpretation of Results

To test whether or not there is some relationship between maintenance cost and the suggested five parameters, this research utilized stochastic analysis step by step. First, through Pearson and Spearman's correlation analysis, correlation and linear relationships could be checked between the dependent variable and the five independent variables. In this process, p-value of the Spearman test was much higher

than the Pearson test in cluster 1 but in cluster 2, the Pearson test result was a better prediction model. In cluster 1, traffic volume was slightly lower than the significance level ($p\text{-value} = 0.069$); for traffic volume (Tv) ranged from 1,000 to 3,500. In cluster 2, Pt had a slightly higher $p\text{-value}$ than the significance level.

Second, through checking $p\text{-value}$ in the ANOVA table for a suggested regression model, relevancy could be checked. The $p\text{-values}$ of the two suggested regression models displayed a high significance of the models.

Third, individual coefficient tests disclosed each variable's significance level when the independent variables were combined in one equation. In this phase, traffic volume and pavement thickness were rejected from the model in cluster 1 since $p\text{-values}$ were more than 0.05 of the significance level. However, in cluster 2, only one independent variable, Pt, was rejected.

The result of a slightly high $p\text{-value}$ of Tv and Pa in the individual coefficient test in cluster 1 can be interpreted in this way: at a low traffic volume level, the regional environment -including annual total rainfall, average temperature, and pavement age- has more pivotal influence on maintenance cost than pavement thickness and traffic volume. That is, traffic volume (Tv) and pavement thickness (Pt) have relatively low influence on pavement under low traffic loading.

The meaning of the suggested prediction model can be explained in this way:

Cluster 1 prediction model: $M_c = \{0.038 * A_t - 0.029 * A_p + 0.156 * P_a\}^{10}$

1. The temperature variable has a positive functional relation to maintenance cost meaning that if temperature increases, costs would also increase. In many studies, temperature has been considered a critical factor impacting the pavement condition (Bosscher et al. 1998; Jia et al. 2007; Yavuzturk et al. 2005).
2. The precipitation in this model has an inverse functional relation to maintenance cost, meaning that if annual total rainfall increases, cost would decrease. This result differed from former studies related to rainfall: much precipitation accelerates the deterioration of pavement condition (Bosscher et al. 1998; Rogers 2003; Saraf et al. 1987), but these studies were conducted only for rainfall. However, this research condition was combined with low traffic condition and a high range of temperatures in Texas provided by data sets. At Table 22, the first year maintenance cost trend combines temperature and precipitation. Basically, this trend is the same for pavement of all ages. This table shows that increased rainfall lowers the cost and a temperature increase causes a cost increase. Temperature and traffic volume influence pavements' physical condition creating material shrink and expansion causing cracks, and precipitation accelerates the pavement damage (Kapiri et al. 2000; Saraf et al. 1987; Wahhab and Balghunaim 1994). However, at a low level of traffic volume, precipitation prevents a cost increase. In other words, at a very low

traffic volume, this research shows that precipitation has a positive effect rather than a negative effect on pavement unlike former engineering studies. Thus, to interpret this result, combined studies reviewing precipitation, temperature and low traffic volume should be conducted.

3. The pavement age variable has a positive functional relationship to maintenance cost, meaning that if pavement age increases, costs will decrease. This result is supported by engineering research results.

Cluster 2 prediction model:

$$Mc = -4404.96 + 41.762 * A_t + 21.698 * A_p + 0.198 * T_v + 7.592 * P_a$$

1. Temperature, precipitation, traffic volume and pavement age have positive functional relations to maintenance cost. This means that if these variables increase, maintenance cost decrease. Many engineering studies support this result.

Table 22. Maintenance cost trend in the first year for cluster 1

Annual Rainfall (inch)	Temperature								
	63	64	65	66	67	68	69	70	71
57	0	1	1	1	2	2	3	5	6
56	0	1	1	2	2	3	4	6	8
55	1		1	2	3	4	6	8	10
54	1		2	3	4	5	7	10	13
53	1			3	5	7	9	12	16
52				4	6	8	11	15	20
51	2				8	10	14	19	25
50	3				10	13	17	23	30
49	3				12	16	22	29	37
48	4					20	27	35	46
47	6	8				25	33	43	55
46	7	10	15			31	40	52	67
45	9	12	17				49	63	81
44	12	15	21				60	76	98
43	14	19	25	33				92	117
42	18	24	31	41				111	140
41	22	29	38	50	60			132	167
40	28	36	47	60	77			158	198
39	34	44	57	73	93			188	235
38	41	54	69	88				223	277
37	50	65	83	106	134	169	211	263	327
36	61	79	100	127	160	200	250	310	384
35	74	95	120	151	190	237	295	365	450
34	89	113	143	180	225	280	347	428	526
33	107	136	171	214	266	330	408	501	613
32	128	162	203	253	314	388	477	585	713
31	153	192	240	298	369	455	557	681	828
30	182	228	284	351	433	531	650	791	959
29	216	269	334	412	506	620	755	917	1109

Temperature: 63°F ~ 71°F

Precipitation: 29 inches ~ 57 inches

Traffic Volume: AADT 1,000 ~ 3,500

2. The R-square value in cluster 1 is much higher than that of cluster 2; this is also an interesting result to be explained with the third result. If low traffic volume represents a low requirement for pavement maintenance, it can be explained that the uncertainty might increase according to increase of traffic volume. In other words, pavement with relatively low traffic volume is mainly influenced by environmental factors which provide stable conditions for the pavement, but in case of high traffic volume, the uncertainty will increase and cause a lower R-square value in cluster 2 than that of cluster 1. In addition, with increased traffic volume, more independent variable were included in the model. In other words, uncertainty originating from a high traffic volume requires more explainable variables to predict exact maintenance cost.

7.2 Recommendation

In the maintenance cost table in the appendix, red box area shows deviational value calculated from the prediction model. One characteristic of the regression model is to decrease its explainable reliability when the prediction intervals are located out of range. Thus, the red boxes mean low reliability for those values. However, the pattern shown in these tables was similar to pavement rehabilitation and surfacing overlay time intervals. Assuming that life expectancy of ACP is about 5 to 10 years, the periods for roadway rehabilitation or overlay are similar to the term in these tables. Therefore, the rehabilitation interval or surface overlay might be predicted

through further study for highway maintenance practices using the suggested five factors and adding other factors.

In addition, the result of the prediction model in cluster 1 shows the possibilities of the opposite impact of precipitation on pavement under very low traffic volume. Thus, further study in regard to this phenomenon should be conducted to explain this trend.

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APPENDIX

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 63.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	216	526	1,188	2,526	5,093	9,807	18,139	32,373	55,974	94,068
30	182	450	1,029	2,211	4,498	8,730	16,259	29,199	50,766	85,741
31	153	384	889	1,931	3,966	7,760	14,557	26,308	45,998	78,082
32	128	327	767	1,684	3,491	6,889	13,017	23,677	41,638	71,045
33	107	277	660	1,466	3,069	6,107	11,625	21,286	37,652	64,584
34	89	235	566	1,273	2,692	5,405	10,369	19,114	34,014	58,657
35	74	198	485	1,103	2,358	4,777	9,236	17,143	30,695	53,223
36	61	167	414	955	2,062	4,215	8,216	15,358	27,670	48,247
37	50	140	353	824	1,800	3,714	7,298	13,741	24,916	43,694
38	41	117	300	710	1,568	3,267	6,474	12,279	22,412	39,531
39	34	98	254	610	1,363	2,869	5,734	10,959	20,136	35,729
40	28	81	215	523	1,183	2,515	5,071	9,768	18,071	32,259
41	22	67	181	447	1,024	2,201	4,479	8,695	16,198	29,095
42	18	55	152	382	885	1,922	3,949	7,729	14,501	26,213
43	14	46	128	325	763	1,676	3,476	6,861	12,967	23,591
44	12	37	107	276	656	1,458	3,055	6,081	11,580	21,207
45	9	30	89	233	563	1,267	2,680	5,382	10,328	19,043
46	7	25	74	197	482	1,098	2,347	4,756	9,199	17,079
47	6	20	61	166	412	950	2,052	4,197	8,182	15,299
48	4	16	50	139	351	820	1,791	3,697	7,268	13,688
49	3	13	41	116	298	706	1,560	3,252	6,447	12,232
50	3	10	34	97	253	607	1,356	2,856	5,710	10,916
51	2	8	27	81	214	520	1,177	2,503	5,050	9,729
52	2	6	22	67	180	445	1,019	2,190	4,459	8,660
53	1	5	18	55	151	379	880	1,913	3,932	7,697
54	1	4	14	45	127	323	759	1,668	3,461	6,832
55	1	3	11	37	106	274	653	1,451	3,041	6,056
56	0	2	9	30	88	232	560	1,261	2,668	5,359
57	0	2	7	25	73	196	480	1,093	2,337	4,736

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 64.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	269	643	1,431	3,001	5,980	11,399	20,896	37,001	63,525	106,081
30	228	552	1,242	2,632	5,292	10,165	18,760	33,419	57,686	96,797
31	192	472	1,076	2,305	4,676	9,052	16,823	30,153	52,334	88,251
32	162	403	931	2,014	4,125	8,050	15,067	27,177	47,433	80,390
33	136	343	803	1,757	3,633	7,149	13,479	24,467	42,949	73,165
34	113	292	692	1,530	3,195	6,340	12,042	22,004	38,851	66,530
35	95	247	594	1,330	2,804	5,615	10,745	19,766	35,107	60,441
36	79	209	509	1,154	2,458	4,964	9,575	17,734	31,692	54,859
37	65	176	435	999	2,150	4,383	8,521	15,893	28,579	49,745
38	54	148	371	863	1,878	3,863	7,572	14,226	25,743	45,064
39	44	124	316	744	1,636	3,400	6,720	12,717	23,164	40,783
40	36	103	268	639	1,424	2,987	5,955	11,355	20,819	36,872
41	29	86	227	549	1,236	2,620	5,269	10,125	18,690	33,302
42	24	71	191	470	1,071	2,294	4,655	9,016	16,759	30,046
43	19	59	161	401	926	2,005	4,107	8,018	15,010	27,079
44	15	48	135	341	799	1,749	3,617	7,120	13,427	24,378
45	12	40	113	290	688	1,523	3,180	6,314	11,995	21,923
46	10	32	94	246	591	1,324	2,792	5,591	10,703	19,692
47	8	26	78	208	506	1,148	2,447	4,943	9,537	17,668
48	6	21	65	175	433	994	2,140	4,364	8,486	15,833
49	5	17	53	147	369	858	1,869	3,846	7,541	14,171
50	4	14	44	123	314	740	1,629	3,385	6,692	12,668
51	3	11	36	103	266	636	1,417	2,974	5,930	11,310
52	2	9	29	85	225	546	1,230	2,608	5,247	10,084
53	2	7	24	71	190	467	1,066	2,283	4,636	8,980
54	1	5	19	58	160	399	921	1,996	4,089	7,985
55	1	4	15	48	134	340	795	1,741	3,601	7,091
56	1	3	12	39	112	288	684	1,516	3,166	6,288
57	1	3	10	32	93	244	588	1,317	2,779	5,567

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 65.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	334	783	1,716	3,554	7,004	13,220	24,026	42,216	71,980	119,456
30	284	674	1,494	3,124	6,209	11,809	21,602	38,181	65,442	109,121
31	240	579	1,298	2,742	5,497	10,534	19,401	34,496	59,444	99,597
32	203	496	1,126	2,402	4,859	9,385	17,404	31,134	53,945	90,828
33	171	424	974	2,101	4,289	8,350	15,594	28,071	48,908	82,759
34	143	361	841	1,834	3,780	7,419	13,955	25,281	44,298	75,342
35	120	307	725	1,598	3,325	6,582	12,472	22,743	40,083	68,529
36	100	260	623	1,390	2,921	5,831	11,133	20,437	36,233	62,275
37	83	220	534	1,206	2,561	5,159	9,925	18,344	32,719	56,540
38	69	186	457	1,045	2,242	4,556	8,836	16,445	29,514	51,284
39	57	156	390	903	1,959	4,018	7,856	14,725	26,595	46,472
40	47	131	332	779	1,708	3,538	6,975	13,169	23,938	42,071
41	38	109	282	670	1,487	3,110	6,184	11,763	21,523	38,048
42	31	91	239	575	1,292	2,729	5,474	10,493	19,329	34,375
43	25	76	202	493	1,120	2,391	4,839	9,348	17,338	31,024
44	21	62	170	421	969	2,091	4,270	8,316	15,534	27,970
45	17	51	142	359	837	1,825	3,763	7,388	13,901	25,189
46	13	42	119	305	721	1,590	3,311	6,555	12,424	22,660
47	11	35	99	259	620	1,383	2,908	5,807	11,090	20,361
48	8	28	83	219	531	1,200	2,549	5,137	9,886	18,275
49	7	23	68	185	455	1,039	2,231	4,537	8,801	16,383
50	5	18	57	155	388	898	1,949	4,001	7,824	14,669
51	4	15	46	130	330	775	1,700	3,522	6,946	13,118
52	3	12	38	109	280	667	1,480	3,096	6,158	11,717
53	2	9	31	90	237	572	1,285	2,717	5,451	10,451
54	2	7	25	75	200	490	1,114	2,380	4,818	9,310
55	1	6	20	62	169	419	964	2,081	4,252	8,283
56	1	5	16	51	142	357	832	1,817	3,747	7,358
57	1	4	13	42	118	304	717	1,583	3,296	6,528

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 66.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	412	950	2,052	4,197	8,182	15,299	27,571	48,083	81,436	134,331
30	351	820	1,791	3,697	7,268	13,688	24,826	43,544	74,126	122,839
31	298	706	1,560	3,252	6,447	12,232	22,330	39,394	67,412	112,239
32	253	607	1,356	2,856	5,710	10,916	20,062	35,604	61,250	102,469
33	214	520	1,177	2,503	5,050	9,729	18,003	32,145	55,600	93,472
34	180	445	1,019	2,190	4,459	8,660	16,136	28,991	50,424	85,191
35	151	379	880	1,913	3,932	7,697	14,446	26,119	45,685	77,577
36	127	323	759	1,668	3,461	6,832	12,916	23,505	41,351	70,581
37	106	274	653	1,451	3,041	6,056	11,534	21,129	37,391	64,158
38	88	232	560	1,261	2,668	5,359	10,287	18,972	33,775	58,266
39	73	196	480	1,093	2,337	4,736	9,162	17,015	30,477	52,866
40	60	165	410	945	2,043	4,179	8,149	15,241	27,472	47,920
41	50	138	349	816	1,783	3,681	7,238	13,636	24,736	43,395
42	41	116	297	702	1,553	3,238	6,420	12,184	22,248	39,258
43	33	96	251	603	1,350	2,843	5,686	10,873	19,987	35,479
44	27	80	213	517	1,171	2,492	5,028	9,690	17,935	32,031
45	22	66	179	442	1,014	2,180	4,440	8,625	16,075	28,887
46	18	55	150	377	876	1,904	3,914	7,666	14,391	26,024
47	14	45	126	321	755	1,660	3,445	6,804	12,866	23,419
48	11	37	105	272	650	1,444	3,028	6,030	11,489	21,051
49	9	30	88	231	557	1,254	2,656	5,337	10,246	18,901
50	7	24	73	195	477	1,087	2,326	4,716	9,125	16,950
51	6	20	60	164	408	940	2,033	4,161	8,116	15,183
52	4	16	49	137	347	811	1,774	3,665	7,209	13,583
53	3	13	41	115	295	699	1,545	3,223	6,393	12,137
54	3	10	33	96	250	600	1,343	2,830	5,662	10,830
55	2	8	27	80	211	515	1,165	2,480	5,007	9,652
56	2	6	22	66	178	440	1,009	2,170	4,421	8,590
57	1	5	18	54	150	375	871	1,895	3,897	7,635

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 67.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	506	1,148	2,447	4,943	9,537	17,668	31,580	54,675	91,994	150,853
30	433	994	2,140	4,364	8,486	15,833	28,476	49,576	83,833	138,090
31	369	858	1,869	3,846	7,541	14,171	25,650	44,910	76,328	126,307
32	314	740	1,629	3,385	6,692	12,668	23,079	40,642	69,434	115,437
33	266	636	1,417	2,974	5,930	11,310	20,742	36,743	63,106	105,416
34	225	546	1,230	2,608	5,247	10,084	18,620	33,184	57,301	96,185
35	190	467	1,066	2,283	4,636	8,980	16,696	29,939	51,982	87,687
36	160	399	921	1,996	4,089	7,985	14,953	26,981	47,111	79,872
37	134	340	795	1,741	3,601	7,091	13,375	24,290	42,655	72,689
38	112	288	684	1,516	3,166	6,288	11,948	21,842	38,581	66,093
39	93	244	588	1,317	2,779	5,567	10,660	19,619	34,862	60,041
40	77	206	504	1,142	2,435	4,922	9,499	17,602	31,468	54,492
41	64	174	431	989	2,130	4,345	8,452	15,773	28,374	49,408
42	53	146	367	854	1,860	3,830	7,511	14,117	25,557	44,756
43	43	122	312	736	1,621	3,370	6,665	12,619	22,995	40,502
44	36	102	265	633	1,410	2,960	5,905	11,266	20,665	36,615
45	29	85	224	543	1,224	2,596	5,225	10,044	18,551	33,067
46	23	70	189	465	1,060	2,273	4,616	8,944	16,633	29,832
47	19	58	159	397	917	1,986	4,071	7,953	14,896	26,884
48	15	48	133	338	791	1,733	3,585	7,062	13,323	24,201
49	12	39	111	287	681	1,508	3,152	6,262	11,902	21,762
50	10	32	93	243	585	1,311	2,767	5,544	10,618	19,546
51	8	26	77	205	501	1,137	2,424	4,901	9,461	17,535
52	6	21	64	173	428	984	2,120	4,326	8,418	15,713
53	5	17	53	145	365	850	1,851	3,813	7,480	14,063
54	4	14	43	121	310	732	1,613	3,355	6,637	12,570
55	3	11	35	101	263	629	1,403	2,947	5,881	11,221
56	2	9	29	84	223	540	1,218	2,584	5,203	10,005
57	2	7	23	70	188	462	1,055	2,262	4,596	8,908

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 68.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	620	1,383	2,908	5,807	11,090	20,361	36,106	62,069	103,769	169,182
30	531	1,200	2,549	5,137	9,886	18,275	32,603	56,351	94,669	155,025
31	455	1,039	2,231	4,537	8,801	16,383	29,409	51,111	86,293	141,943
32	388	898	1,949	4,001	7,824	14,669	26,499	46,314	78,590	129,863
33	330	775	1,700	3,522	6,946	13,118	23,851	41,926	71,511	118,716
34	280	667	1,480	3,096	6,158	11,717	21,443	37,916	65,012	108,438
35	237	572	1,285	2,717	5,451	10,451	19,257	34,254	59,049	98,968
36	200	490	1,114	2,380	4,818	9,310	17,273	30,914	53,583	90,249
37	169	419	964	2,081	4,252	8,283	15,475	27,870	48,577	82,228
38	142	357	832	1,817	3,747	7,358	13,848	25,098	43,995	74,854
39	118	304	717	1,583	3,296	6,528	12,376	22,577	39,806	68,080
40	99	257	616	1,376	2,894	5,783	11,046	20,286	35,980	61,863
41	82	218	529	1,194	2,538	5,115	9,846	18,207	32,488	56,162
42	68	183	452	1,034	2,221	4,517	8,765	16,321	29,304	50,938
43	56	154	386	894	1,940	3,983	7,792	14,613	26,403	46,156
44	46	129	328	771	1,692	3,507	6,917	13,068	23,764	41,781
45	38	108	279	663	1,473	3,082	6,132	11,671	21,365	37,784
46	31	90	236	569	1,279	2,705	5,428	10,410	19,185	34,134
47	25	75	199	488	1,109	2,369	4,797	9,273	17,208	30,804
48	20	62	168	417	959	2,072	4,234	8,249	15,416	27,770
49	16	51	141	355	828	1,808	3,730	7,328	13,794	25,007
50	13	42	118	302	713	1,575	3,281	6,501	12,327	22,494
51	10	34	98	256	613	1,370	2,881	5,758	11,003	20,211
52	8	28	82	216	526	1,188	2,526	5,093	9,807	18,139
53	7	22	68	182	450	1,029	2,211	4,498	8,730	16,259
54	5	18	56	153	384	889	1,931	3,966	7,760	14,557
55	4	15	46	128	327	767	1,684	3,491	6,889	13,017
56	3	12	38	107	277	660	1,466	3,069	6,107	11,625
57	2	9	31	89	235	566	1,273	2,692	5,405	10,369

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 69.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	755	1,660	3,445	6,804	12,866	23,419	41,208	70,351	116,884	189,491
30	650	1,444	3,028	6,030	11,489	21,051	37,260	63,947	106,750	173,806
31	557	1,254	2,656	5,337	10,246	18,901	33,656	58,072	97,413	159,299
32	477	1,087	2,326	4,716	9,125	16,950	30,369	52,688	88,818	145,892
33	408	940	2,033	4,161	8,116	15,183	27,373	47,757	80,911	133,509
34	347	811	1,774	3,665	7,209	13,583	24,646	43,246	73,644	122,080
35	295	699	1,545	3,223	6,393	12,137	22,166	39,122	66,970	111,539
36	250	600	1,343	2,830	5,662	10,830	19,913	35,355	60,845	101,824
37	211	515	1,165	2,480	5,007	9,652	17,868	31,918	55,228	92,878
38	178	440	1,009	2,170	4,421	8,590	16,014	28,784	50,083	84,646
39	150	375	871	1,895	3,897	7,635	14,335	25,930	45,373	77,076
40	125	319	751	1,652	3,430	6,776	12,817	23,334	41,066	70,121
41	105	271	646	1,437	3,014	6,005	11,444	20,974	37,130	63,736
42	87	229	554	1,248	2,644	5,314	10,205	18,830	33,537	57,879
43	72	193	475	1,082	2,315	4,696	9,089	16,886	30,261	52,511
44	60	163	405	935	2,024	4,143	8,083	15,125	27,275	47,595
45	49	137	345	807	1,766	3,649	7,179	13,531	24,557	43,097
46	40	114	293	695	1,538	3,209	6,367	12,089	22,085	38,986
47	33	95	249	597	1,337	2,817	5,638	10,788	19,839	35,231
48	27	79	210	512	1,159	2,469	4,986	9,613	17,801	31,805
49	22	65	177	438	1,004	2,160	4,402	8,555	15,954	28,681
50	17	54	149	373	867	1,886	3,880	7,603	14,280	25,837
51	14	44	124	318	747	1,644	3,415	6,748	12,767	23,249
52	11	36	104	269	643	1,431	3,001	5,980	11,399	20,896
53	9	30	86	228	552	1,242	2,632	5,292	10,165	18,760
54	7	24	72	192	472	1,076	2,305	4,676	9,052	16,823
55	6	19	59	162	403	931	2,014	4,125	8,050	15,067
56	4	16	49	136	343	803	1,757	3,633	7,149	13,479
57	3	12	40	113	292	692	1,530	3,195	6,340	12,042

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 70.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	917	1,986	4,071	7,953	14,896	26,884	46,950	79,614	131,472	211,968
30	791	1,733	3,585	7,062	13,323	24,201	42,508	72,452	120,201	194,611
31	681	1,508	3,152	6,262	11,902	21,762	38,447	65,876	109,807	178,544
32	585	1,311	2,767	5,544	10,618	19,546	34,739	59,841	100,229	163,680
33	501	1,137	2,424	4,901	9,461	17,535	31,356	54,309	91,409	149,940
34	428	984	2,120	4,326	8,418	15,713	28,273	49,241	83,294	137,247
35	365	850	1,851	3,813	7,480	14,063	25,465	44,603	75,834	125,529
36	310	732	1,613	3,355	6,637	12,570	22,911	40,362	68,980	114,719
37	263	629	1,403	2,947	5,881	11,221	20,589	36,487	62,689	104,754
38	223	540	1,218	2,584	5,203	10,005	18,482	32,951	56,919	95,576
39	188	462	1,055	2,262	4,596	8,908	16,570	29,726	51,632	87,127
40	158	394	912	1,977	4,053	7,920	14,839	26,788	46,791	79,357
41	132	336	787	1,724	3,569	7,033	13,272	24,113	42,362	72,216
42	111	285	677	1,501	3,138	6,235	11,855	21,682	38,314	65,659
43	92	242	582	1,304	2,754	5,521	10,576	19,473	34,618	59,642
44	76	204	498	1,131	2,413	4,880	9,423	17,470	31,245	54,126
45	63	172	426	979	2,110	4,308	8,384	15,653	28,172	49,074
46	52	144	363	845	1,842	3,796	7,449	14,008	25,373	44,450
47	43	121	309	728	1,606	3,340	6,610	12,521	22,827	40,222
48	35	101	262	626	1,396	2,934	5,856	11,177	20,513	36,360
49	29	84	221	537	1,212	2,573	5,181	9,965	18,412	32,834
50	23	69	187	460	1,050	2,252	4,576	8,872	16,508	29,620
51	19	57	157	392	908	1,968	4,036	7,888	14,782	26,691
52	15	47	132	334	783	1,716	3,554	7,004	13,220	24,026
53	12	39	110	284	674	1,494	3,124	6,209	11,809	21,602
54	10	32	92	240	579	1,298	2,742	5,497	10,534	19,401
55	8	26	76	203	496	1,126	2,402	4,859	9,385	17,404
56	6	21	63	171	424	974	2,101	4,289	8,350	15,594
57	5	17	52	143	361	841	1,834	3,780	7,419	13,955

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

Highway Pavement Maintenance Cost Table

AADT range from 1,000 to 3,500

Pavement thickness ignored

Annual Average Temperature: 71.0 °F

Cost: Dollar per lane mile per year

Annual Precipitation (inch)	Pavement Age from last overlay date (year)									
	1	2	3	4	5	6	7	8	9	10
29	1,109	2,369	4,797	9,273	17,208	30,804	53,403	89,961	147,679	236,817
30	959	2,072	4,234	8,249	15,416	27,770	48,412	81,963	135,159	217,631
31	828	1,808	3,730	7,328	13,794	25,007	43,844	74,610	123,602	199,855
32	713	1,575	3,281	6,501	12,327	22,494	39,669	67,857	112,942	183,397
33	613	1,370	2,881	5,758	11,003	20,211	35,854	61,658	103,117	168,169
34	526	1,188	2,526	5,093	9,807	18,139	32,373	55,974	94,068	154,089
35	450	1,029	2,211	4,498	8,730	16,259	29,199	50,766	85,741	141,078
36	384	889	1,931	3,966	7,760	14,557	26,308	45,998	78,082	129,065
37	327	767	1,684	3,491	6,889	13,017	23,677	41,638	71,045	117,980
38	277	660	1,466	3,069	6,107	11,625	21,286	37,652	64,584	107,760
39	235	566	1,273	2,692	5,405	10,369	19,114	34,014	58,657	98,344
40	198	485	1,103	2,358	4,777	9,236	17,143	30,695	53,223	89,674
41	167	414	955	2,062	4,215	8,216	15,358	27,670	48,247	81,699
42	140	353	824	1,800	3,714	7,298	13,741	24,916	43,694	74,368
43	117	300	710	1,568	3,267	6,474	12,279	22,412	39,531	67,634
44	98	254	610	1,363	2,869	5,734	10,959	20,136	35,729	61,454
45	81	215	523	1,183	2,515	5,071	9,768	18,071	32,259	55,787
46	67	181	447	1,024	2,201	4,479	8,695	16,198	29,095	50,595
47	55	152	382	885	1,922	3,949	7,729	14,501	26,213	45,841
48	46	128	325	763	1,676	3,476	6,861	12,967	23,591	41,494
49	37	107	276	656	1,458	3,055	6,081	11,580	21,207	37,521
50	30	89	233	563	1,267	2,680	5,382	10,328	19,043	33,894
51	25	74	197	482	1,098	2,347	4,756	9,199	17,079	30,586
52	20	61	166	412	950	2,052	4,197	8,182	15,299	27,571
53	16	50	139	351	820	1,791	3,697	7,268	13,688	24,826
54	13	41	116	298	706	1,560	3,252	6,447	12,232	22,330
55	10	34	97	253	607	1,356	2,856	5,710	10,916	20,062
56	8	27	81	214	520	1,177	2,503	5,050	9,729	18,003
57	6	22	67	180	445	1,019	2,190	4,459	8,660	16,136

* The cost in red box is out of range of pavement maintenance prediction model

** Base year: 2010

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