# Parametric Analysis of Resilient Modulus Modeling for Recycled Asphalt Pavement in Base Layer 

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# Parametric Analysis of Resilient Modulus Modeling for Recycled Asphalt Pavement in Base Layer 

Ehab M. Noureldin and Magdy Abdelrahman


#### Abstract

The resilient modulus ( $M_{R}$ ) has been found to be the most important parameter in the Mechanistic-Empirical Pavement Design Guide (MEPDG) for the base layer. A literature review showed that the use of recycled asphalt pavement (RAP) in the base layer had many economic and structural benefits. The prediction of $M_{R}$ with the specific MEPDG model for the base layer mixed with RAP was found to be reliable for several field conditions. However, further studies of the MEPDG model are needed for an understanding of the physical meaning of each parameter in the model. The most important factor affecting $M_{R}$ for the base layer is the state of stresses, especially the confining pressure. This type of state of stresses was found in previous research to be more effective than deviator stresses. The present study focuses on the parametric analysis of each constant in the model versus various confining pressure states under various field conditions and at various RAP concentrations. This parametric analysis is compared with the traditional granular coarse aggregates used for the base layer before RAP is used so that any difference in the physical meaning when RAP is mixed in the base layer blend can be determined. The prediction of $M_{R}$ by the MEPDG model appears to be sound for RAP, but only with confining pressure levels below 10 psi and without significant variation in water content and maximum dry density. Freeze-thaw cycles do not negatively influence the prediction of $M_{R}$ for both RAP and granular coarse aggregates.


The resilient modulus ( $M_{R}$ ) of unbound layers is a required property during any mechanistic or mechanistic-empirical analysis for flexible pavements. The $M_{R}$ test is commonly conducted in the laboratory to characterize the stiffness and elasticity responses of the base material (l). Proper characterization of the nonlinear, stress-dependent behavior of pavement unbound layers has a significant effect on the accuracy of pavement response predictions (2). $M_{R}$ of unbound granular material is affected by several factors, including state of stress, moisture content, dry density, and freeze-thaw action. Researchers have sought to understand and model the effect of these factors on $M_{R}$ for various granular materials and to understand the effect of those factors on $M_{R}$ of recycled asphalt pavement (RAP) and RAP-aggregate blends as a pavement base course layer (3).

A literature survey of $M_{R}$ for base layers found that $M_{R}$ depends more on confining pressure than deviator stress levels (1). Several constitutive $M_{R}$ models have been studied for the base layer, especially

[^0]when RAP is used. The Mechanistic-Empirical Pavement Design Guide (MEPDG) model was found to be the best suitable prediction model for $M_{R}$ (4). However, more parametric analysis is needed for the constants $(K)$ of the model as related to state of stresses to gain an understanding of the physical meaning of each parameter. Therefore, this study focuses on the parametric analysis for each $K$ parameter in the MEPDG model for various levels of confining pressure at the same field conditions usually studied for RAP.

## PROBLEM STATEMENT

The MEPDG model is the most reliable model and is the best fit for predicting $M_{R}$ of a granular base layer and when RAP is used as a base layer in various field conditions (3). However, it has been found that there is no exact relationship between each $K$ parameter and the studied field conditions individually (4). These field conditions include various water contents, decreased dry density, and applied freeze-thaw cycles at various RAP concentrations. Therefore, there is a need to assess the physical meaning of each constant in the model, especially for the state of stresses, which is the most influential parameter for $M_{R}$ for the base layer. Also, there is a need for a further study of model suitability for traditional base coarse aggregates and RAP-aggregate mixes, depending on the behavior of each constant parameter of the MEPDG model at different states of stress.

## SCOPE OF WORK

In this model, multiple regression analysis is achieved for the MEPDG model with the measured $M_{R}$ values for various measured testing parameters. In the first stage, each $K$ parameter is calculated for five levels of confining pressure ( $3,6,10,15$, and 20 psi ) with the Excel Solver tool. Six replicates are considered for each level of confining pressure. Each replicate is assigned a value of deviator stress. This stage of the analysis is an addition to a previous study that did not consider the effect of various states of stress on $K$ regression parameters (4). This analysis cannot be conducted at various levels of deviator stress because testing sequences required by the NCHRP 1-28A protocol in $M_{R}$ testing do not group deviator stress as they do confining pressure levels. This regression is repeated under four field conditions:

1. Percentage of RAP $(0 \%, 50 \%, 75 \%$, and $100 \%)$;
2. Six moisture contents, ranging from $\mathrm{OMC}-3 \%$ to $\mathrm{OMC}+2 \%$ with $1 \%$ increments, where OMC is optimum moisture content;
3. Two levels of compaction, $100 \%$ and $90 \%$ maximum dry density (MDD); and
4. Two freeze-thaw cycles.

The MEPDG model follows:
$M_{R}=K_{1} \cdot P_{a}\left(\frac{\theta}{P_{a}}\right)^{K_{2}}\left(\frac{\tau_{\text {oct }}}{P_{a}}+1\right)^{K_{3}}$
where

$$
\begin{aligned}
\theta=\sigma_{1} & +\sigma_{2}+\sigma_{3} \\
& =\sigma_{d}+3 \sigma_{3} \\
& =\text { bulk stress }(\mathrm{psi}), \\
\tau_{\text {oct }}= & \frac{1}{3} \cdot \sqrt{\left\{\left(\sigma_{1}-\sigma_{2}\right)^{2}+\left(\sigma_{2}-\sigma_{3}\right)^{2}+\left(\sigma_{1}-\sigma_{3}\right)^{2}\right\}} \\
& =\text { octahedral shear stress }(\mathrm{psi}), \\
M_{R} & =\text { resilient modulus (psi) }, \\
K_{i} & =\text { multiple regression constants evaluated from } M_{R} \text { tests, } \\
P_{a} & =\text { atmospheric pressure, } 14.7 \text { psi }(101.5 \mathrm{kPa}), \\
\sigma_{d} & =\text { deviator stress (psi), and } \\
\sigma_{3} & =\text { confining pressure (psi). }
\end{aligned}
$$

## LITERATURE REVIEW

The proper modulus for an unbound base course is necessary for good pavement performance. $M_{R}$ is a commonly used parameter for defining material stiffness. It is similar to Young's modulus and is based on the recoverable axial strain $\left(\varepsilon_{r}\right)$ under an imposed axial (deviator) stress $\left(\sigma_{d}\right)$ :
$M_{R}=\frac{\sigma_{d}}{\varepsilon_{r}}$
The Guide for Mechanistic-Empirical Design of New and Rehabilitated Pavement Structures (5) and the 1993 AASHTO Guide for Design of Pavement Structures (6) recommended the use of the resilient modulus of base materials as a material property for characterizing pavements during structural analysis and design (7).

In the laboratory $M_{R}$ test, constant confining pressure is maintained within a conventional triaxial cell, and a cyclic axial stress is applied to simulate traffic loading. The NCHRP 1-28A test protocol used in this study consists of 30 loading sequences, but the protocol loading involves a conditioning stage, which attempts to establish steady state or resilient behavior, through the application of 1,000 cycles of 30 psi deviator stresses at 15 psi confining pressure. The cycles are then repeated 100 times for 30 loading sequences with various combinations of deviator stress and confining pressure. $M_{R}$ is calculated as the mean of the last five cycles of each sequence from the recoverable axial strain and cyclic axial stress (8).

The $M_{R}$ prediction model is called mechanistic-empirical because of the mechanistic calculation of stresses, strains, and deflections of a pavement structure. These properties are the fundamental pavement responses under repeated traffic loadings. The relationship of these responses to field distresses and performance is determined with existing empirical relationships, widely known as transfer functions. The design process is an iterative procedure that starts with a trial design and ends when predicted distresses meet acceptable limits based on the desired level of statistical reliability (9).
$M_{R}$ of base materials can be estimated with laboratory repeated load triaxial tests. It also can be estimated from empirical correlations with soil properties or nondestructive test results. Several researchers have proposed correlations of $M_{R}$ with the stress state (7).

The influence of stress state on unbound material stiffness has long been recognized in pavement engineering. For coarse-grained
granular soils as the base layer, an increase in stiffness with increasing confining stress is usually the dominant effect. Most soils exhibit both effects of increasing stiffness with increasing confinement and decreasing stiffness with increasing shear (10). Witczak and Uzan have proposed a universal model that combines both effects into a single equation (11):
$M_{R}=K_{1} \cdot P_{a}\left(\frac{\theta}{P_{a}}\right)^{K_{2}}\left(\frac{\sigma_{d}}{P_{a}}\right)^{K_{3}}$
where $K_{1}>0, K_{2} \geq 0$, and $K_{3} \leq 0$. A modification of this equation was adopted for the new national pavement design guide being finalized in NCHRP Project 1-37A (12):
$M_{R}=K_{1} \cdot P_{a}\left(\frac{\theta}{P_{a}}\right)^{K_{2}}\left(\frac{\tau_{\text {oct }}}{P_{a}}+1\right)^{K_{3}}$
where 1 in the $\tau_{\text {oct }}$ term is used to avoid numerical problems when $\tau_{\text {oct }}$ approaches zero. The model presented in Equation 4 shows a clear improvement for some materials and tests, but for others model accuracy is diminished. Trial and error in the NCHRP 1-28A project showed that adding $K_{4}$ (as in Equation 5) to the previous model increases accuracy generally, but the model remains inconsistent. For some materials improvement is shown, and for some others the model is less accurate. Because the ( +1 ) model did not consistently show improvement in all tests or for all materials, 1 clearly was not always the right value. Hence, in the last model, as presented in Equation 5, a regression constant, $K_{5}$, was introduced instead of +1 . As expected, the analysis showed that the model exhibits the best overall goodness-of-fit statistics (10):
$M_{R}=K_{1} \cdot P_{a}\left(\frac{\theta-3 K_{4}}{P_{a}}\right)^{K_{2}}\left(\frac{\tau_{\text {oct }}}{P_{a}}+K_{5}\right)^{K_{3}}$
$M_{R}$ is a primary input to any mechanistically oriented pavement design procedure. The pavement design computer code must be adapted to support the specific predictive equation selected for $M_{R}$. NCHRP Project 1-28A made no recommendation about the final form for the $M_{R}$ model. Selection of the final model must be coordinated with the pavement design method or code in which it is to be implemented. NCHRP Project 1-37A, Development of the 2002 Guide for the Design of New and Rehabilitated Pavement Structures, was just getting underway when NCHRP Project 1-28A was being completed.
After a thorough review of the results of Andrei et al. (10) and other studies dealing with the prediction of the resilient response of unbound materials, the NCHRP Project 1-37A team selected the model presented in Equation 4 as the recommended $M_{R}$ model to be implemented and used in the 2002 design guide. This model was deemed the best compromise for accuracy, ease of implementation, and computational stability (for the case of $\tau_{\text {oct }}=0$ ). The model given in Equation 4 was implemented in the 2002 design guide software.

## EXPERIMENTAL CONSIDERATIONS

This study was conducted on one source of RAP collected by the Minnesota Department of Transportation (DOT) from a trunk highway. Use of one source avoids the problem of variability in RAP from several sources or sites. Also, this RAP was used because it contains all data needed in the study for requested field conditions. This material was mixed with Class 5 base aggregates (Minnesota DOT) at $50 \%, 75 \%$, and $100 \%$ RAP. For sample homogeneity, the maximum
particle size should be less than $10 \%$ of the mold size. Therefore, all materials greater than 12.5 mm were replaced by others passing the $12.5-\mathrm{mm}$ sieve and retained on the No. $4(4.75-\mathrm{mm})$ sieve. This sizing was the only adjustment made for aggregate gradation in this stage of the research (13). For this kind of RAP, the OMC is $5.5 \%$ and the MDD is $2,124 \mathrm{~kg} / \mathrm{m}^{3}$; for Class 5 the OMC is $6.4 \%$ and the MDD is $2,223 \mathrm{~kg} / \mathrm{m}^{3}(1)$. For $50 \%$ RAP, the OMC is $5.2 \%$ and the MDD is $2,188 \mathrm{~kg} / \mathrm{m}^{3}$. For $75 \%$ RAP, the OMC is $5.7 \%$ and the MDD is $2,161 \mathrm{~kg} / \mathrm{m}^{3}(1)$. For both $50 \%$ and $75 \%$ RAP blends, weights were calculated according to these percentages and the gradation used was that for normal aggregates used for base layers at each sieve size.

The $M_{R}$ test was conducted immediately after sample compaction. The target sample size was 6 in. in diameter and 12 in . in height. The sample was subjected to 1,000 load cycles for preconditioning, followed by the 30 load sequences specified by the NCHRP 1-28A protocol, Procedure 1A (14). The $M_{R}$ test was conducted inside a triaxial pressure chamber, which maintained the required confining pressure by applying pneumatic pressure with a suitable air compressor. These results of the measured $M_{R}$ previously had been achieved by the North Dakota State University research team with the cooperation of the Minnesota DOT, which supplied the material samples. The samples were also subjected to two freeze-thaw cycles before testing for determining the $M_{R}$ values. There was no specification standard for investigating the freeze-thaw effect on $M_{R}$. Various methods for evaluating different construction materials were reviewed. This review showed that two cycles are enough to cause most of the detrimental effects for the base layer. More than two cycles could be dangerous in cold climates such as North Dakota, and this could be investigated in future research. Samples were frozen at $-12^{\circ} \mathrm{F}$ for 24 h , then thawed for 24 h at $75^{\circ} \mathrm{F}$ (3). Some samples failed during the $M_{R}$ test, and no measured data were available for comparison with predicted values from the MEPDG model.

## ANALYSIS OF RESULTS

A review of the literature on the $M_{R}$ MEPDG model showed that $K_{1}>0$, which refers to the Young's modulus of the material. Also, $K_{2}>0$ refers to stress stiffening, and $K_{3}<0$ refers to shear softening (10). Most of the collected results satisfy the trends drawn from the literature for the tested confining pressure levels and field conditions, which included water content, dry density, and freeze-thaw cycles at various RAP percentages in the blend. Analysis of each $K$ parameter in the model at the various testing conditions is presented next.

## Analysis of $K_{1}$

The variation of the $K_{1}$ parameter is high for variations in confining pressure in the testing conditions. Therefore, the data collected for this parameter is shown in semilog charts in Figures 1, 2, and 3 to represent $K_{1}$ under the variations in the confining pressure factor for all testing conditions at different RAP concentrations. Confining pressure levels can be divided into three categories: low ( 3 and 6 psi ), intermediate ( 10 psi ), and high ( 15 and 20 psi ). Under the variation of water content at different levels of confining pressure and different percentages of RAP (Figure 1), $K_{1}$ increases with increasing percentage of RAP exceeding the original values before use of RAP $(0 \%)$. This relationship is obvious at low confining pressure levels ( 3 and 6 psi ) and most water content levels except at $\mathrm{OMC}-3 \%$.

At the intermediate confining pressure level ( 10 psi ), $K_{1}$ decreases dramatically, approaching zero on granular coarse aggregates before
use of RAP $(0 \%)$. For RAP used at varying water content levels, $K_{1}$ also approaches zero, except at $100 \%$ RAP and water content levels ranging from OMC $-2 \%$ to OMC .

At high confining pressure levels ( 15 and 20 psi ), $K_{1}$ increases dramatically at $0 \%$ RAP, as it does when RAP is used. However, it does not exceed the original values at $0 \%$ RAP. There is an exception for this relationship at low water content levels OMC $-2 \%$ and OMC $-3 \%$ as $K_{1}$ does not increase dramatically like the other water content levels and it decreases with increasing percentage of RAP.

When MDD decreases from $100 \%$ to $90 \%$ (Figure 2), at low confining pressure, $K_{1}$ increases with an increasing percentage of RAP exceeding the original values of $K_{1}$ at granular base coarse aggregates ( $0 \%$ RAP). At intermediate confining pressure, $K_{1}$ decreases dramatically, approaching zero for all percentages of RAP. At high confining pressure levels, $K_{1}$ increases dramatically for all percentages of RAP. In general, $K_{1}$ values at $90 \%$ MDD are higher than $100 \%$ MDD in the other equivalent testing conditions.

For the last case of freeze-thaw cycles (Figure 3) at low confining pressure levels, $K_{1}$ increases with an increasing percentage of RAP exceeding the original $K_{1}$ values at $0 \%$ RAP. This relationship is clearer at the 3 -psi confining pressure than at 6 psi. At intermediate confining pressure, $K_{1}$ decreases, approaching zero, except at $100 \%$ RAP in $\mathrm{OMC}+2 \%$. At high confining pressure, $K_{1}$ increases dramatically except at a high percentage of RAP and water content concentration. Generally, $K_{1}$ values are not affected significantly by freeze-thaw cycles at the same testing field conditions.

## Analysis of $K_{2}$

For all testing conditions, $K_{2}$ is positive at most confining pressure levels and reaches its maximum values for most of the testing conditions at a confining pressure of 10 psi . Therefore, $K_{2}$ appears to be more reliable at an intermediate confining pressure than others. Figures 4,5, and 6 show that $K_{2}$ values are very close to zero and are sometimes negative. This result satisfies results from the literature for this model and confirm that $K_{2}$ values should be positive for the base layer. This analysis describes the behavior of stress stiffening (10).

For water content variation (Figure 4), the maximum values of $K_{2}$ are usually reached when water content is far from OMC (at OMC $-3 \%$ and $\mathrm{OMC}+2 \%$ ), especially at $50 \%$ and $100 \%$ of RAP, where $K_{2}$ values at OMC are lower than $0 \%$ and $75 \%$ of RAP. However, this relationship is not the same for granular base aggregates as shown at $0 \%$ RAP; the maximum values reached at OMC $-1 \%$ and $\mathrm{OMC}+1 \%$. The comparison of RAP concentrations and granular base aggregates shows that the trend of $K_{2}$ appears to be very close at $75 \%$ RAP. Generally, maximum $K_{2}$ values were reached in water contents far from OMC and ranged between 50 and 80 at $50 \%$ RAP, between 70 and 140 at $75 \%$ RAP, and around 50 at $100 \%$ RAP. For granular base aggregates, values range between 75 and 100 at water contents close to OMC.

When MDD is decreased to $90 \%$ (Figure 5), $K_{2}$ values increase, reaching 170 for $50 \%$ and $100 \%$ RAP and 250 for $0 \%$ RAP. These optimum values are reached especially with high water content levels, such as OMC $+2 \%$. $K_{2}$ values at OMC $-2 \%$ for $50 \%$ and $100 \%$ RAP significantly decrease and reach nonrealistic negative values for $K_{2}$. This behavior is not the same as that for granular base aggregates ( $0 \% \mathrm{RAP}$ ) at high water content.

After freeze-thaw cycles (Figure 6) were applied at the different water contents, $K_{2}$ values ranged between 65 and 75, especially at OMC $+2 \%$ for $50 \%$ and $75 \%$ RAP. However, at $100 \%$ RAP, $K_{2}$ values decrease to 55 at $\mathrm{OMC}+1 \%$. For granular base aggregates

(a)


(b)

(d)

| $-\cdots$ OMC - 3 | $-\cdots-$ OMC - | $-=-$ OMC - 1 |
| :--- | :--- | :--- |
| $\cdots$ OMC | $\cdots \cdots-\cdots$ OMC +1 | $\cdots-\cdots$ OMC +2 |

FIGURE $1 \quad K_{1}$ versus water content variation: water content effect at (a) 0\% RAP, (b) 50\% RAP, (c) $75 \%$ RAP, and (d) 100\% RAP (conf. = confining).
( $0 \% \mathrm{RAP}$ ), optimum $K_{2}$ values reach 80 at $\mathrm{OMC}+2 \%$ and 65 at OMC. This is much higher than for $50 \%$ and $100 \%$ RAP. Application of freeze-thaw cycles affects $K_{2}$ values positively as it increases after freeze-thaw cycles at the same water content levels for all percentages of RAP.

## Analysis of $K_{3}$

As shown in Figures 7, 8, and 9 for all testing conditions, $K_{3}$ is negative at most confining pressure levels and reaches its minimum values for most of the testing conditions at a confining pressure of 10 psi . This result satisfies those in the literature that state $K_{3}$ values should be negative for the base layer. This analysis describes the behavior of shear softening (10).

In the case of water content variation (Figure 7), $K_{3}$ values reach their minimum values at water content levels away from OMC, such as $\mathrm{OMC}-3 \%$ and $\mathrm{OMC}+2 \%$. This trend is obvious at $50 \%, 75 \%$, and $100 \%$ of RAP. The lowest $K_{3}$ values are approximately $-80,-140$,
and -60 for cases of $50 \%, 75 \%$, and $100 \%$ RAP, respectively. At $0 \%$ RAP, the lowest $K_{3}$ value reaches -110 at a water content level of OMC $-1 \%$. A comparison of $0 \%$ and $100 \%$ RAP suggests that the general trend of $K_{3}$ values at $0 \%$ is much lower than $100 \%$, which means that this model is more suitable for granular base aggregates than RAP.

When MDD is decreased to $90 \%$ (Figure 8 ), $K_{3}$ values are lower than $100 \%$ MDD, reaching -170 at OMC $+2 \%$ for $50 \%$ and $100 \%$ RAP. In the case of $0 \% \mathrm{RAP}, K_{3}$ values are lower than those for all percentages of RAP $(50 \%$ and $100 \%)$ at the three water contents of $\mathrm{OMC}-2 \%, \mathrm{OMC}$, and $\mathrm{OMC}+2 \%$. At $0 \% \mathrm{RAP}, K_{3}$ reaches -250 in the case of OMC $+2 \%$, compared with -170 at RAP $50 \%$ and $100 \%$. This result matches the conclusion in the previous case of water content variation.

After freeze-thaw cycles (Figure 9) at the different water contents are applied, $K_{3}$ reaches -80 at OMC $+2 \%$ for $50 \%$ RAP, compared with -60 in the same field conditions before freeze-thaw cycles. $K_{3}$ reaches -65 at OMC $+2 \%$ for $75 \%$ RAP and reaches -50 at $100 \%$ RAP and $\mathrm{OMC}+1 \%$. For the case of granular coarse aggregates $(0 \% \mathrm{RAP}), K_{3}$ reaches -80 at $\mathrm{OMC}+2 \%$, compared with -50 at

(a)

(b)

(c)

$$
\begin{array}{|cc|c|}
\hline-\cdots \mathrm{OMC}-2 \quad \longrightarrow \text { OMC } \quad \cdots \cdot \mathrm{OMC}+2 \\
\hline
\end{array}
$$

FIGURE $2 \quad K_{1}$ versus MDD variation: 90\% MDD effect at (a) 0\% RAP, (b) 50\% RAP, and (c) 100\% RAP.

(a)

Conf. Pressure (psi)

(c)

(b)

Conf. Pressure (psi)

(d)

$$
\begin{array}{|lllll}
\hline \leftrightarrow & \cdots & \text { OMC } & \cdots \cdots & \text { OMC }+1 \\
\cdots & \text { OMC }+2 \\
\hline
\end{array}
$$

FIGURE $3 \quad K_{1}$ versus freeze-thaw cycles: effect at (a) 0\% RAP, (b) 50\% RAP, (c) 75\% RAP, and (d) 100\% RAP.

(a)

(c)

(b)

(d)

| -*- OMC - 3 | - - OMC-2 | - - OMC - 1 |
| :---: | :---: | :---: |
| $\cdots$ OMC | ---*--- OMC + 1 | -... OMC + 2 |

FIGURE $4 K_{2}$ versus water content variation: water content effect at (a) 0\% RAP, (b) 50\% RAP, (c) $75 \%$ RAP, and (d) $100 \%$ RAP.


- OMC-2 $\quad$ OMC $\quad \cdots \quad$ OMC + 2

FIGURE $5 K_{2}$ versus MDD variation: 90\% MDD effect at (a) 0\% RAP, (b) $50 \%$ RAP, and (c) 100\% RAP.

(a)

(c)

(b)

(d)


FIGURE $6 K_{2}$ versus freeze-thaw cycles: effect at (a) 0\% RAP, (b) 50\% RAP, (c) 75\% RAP, and (d) 100\% RAP.
Conf. Pressure (psi)

(a)
(c)

(b)

(d)

| $\begin{aligned} & -* \text { OMC }-3 \\ & - \text { OMC } \end{aligned}$ | $\begin{aligned} & -- \text { OMC - } 2 \\ & \cdots-\cdots \text { OMC + } 1 \end{aligned}$ | $\begin{aligned} & -=-\mathrm{OMC}-1 \\ & \cdots-\mathrm{OMC}+2 \end{aligned}$ |
| :---: | :---: | :---: |

FIGURE $7 K_{3}$ versus water content variation: water content effect at (a) 0\% RAP, (b) 50\% RAP, (c) 75\% RAP, and (d) 100\% RAP.


FIGURE $8 \quad K_{3}$ versus MDD variation: 90\% MDD effect at (a) 0\% RAP, (b) $50 \%$ RAP, and (c) $100 \%$ RAP.

(a)

(b)

(c)

(d)

$$
\square \text { OMC } \quad \cdots \cdots \text { OMC + } 1 \quad \cdots \cdots \cdots \text { OMC + } 2
$$

FIGURE $9 K_{3}$ versus freeze-thaw cycles: effect at (a) 0\% RAP, (b) 50\% RAP, (c) 75\% RAP, and (d) 100\% RAP.
the same case before freeze-thaw cycles. Thus, freeze-thaw cycles affect more positively $K_{3}$ absolute values. A comparison of granular coarse aggregates ( $0 \%$ ) and RAP ( $100 \%$ ) shows that traditional coarse aggregates are more suitable to the model as $K_{3}$ absolute values are higher, and this describes shear softening behavior for a base layer better than RAP.

## Analysis of RAP-Aggregate Combination

After analysis of each $K$ parameter with various confining pressure levels under various testing conditions, the variation of each $K$ parameter was assessed with different RAP concentrations for all confining pressure levels previously studied. In this case, testing conditions are concerned only with OMC and MDD without freezethaw cycles because these are the most frequent field conditions likely to happen early in the life of a pavement section. This stage of analysis is achieved to differentiate between the behaviors of each $K$ parameter when granular base coarse aggregates or RAP is used.

The following analyses are shown in Figure 10. $K_{1}$ values are low for granular base coarse aggregates and RAP concentrations ( $50 \%$, $75 \%$, and $100 \%$ ) at low and intermediate confining pressure levels
(Figure $10 a$ ). However, $K_{1}$ increases dramatically at high confining pressure levels ( 15 and 20 psi ), especially for base coarse aggregates. It increases significantly with increasing RAP concentration.
$K_{2}$ values are almost zero for all RAP concentrations (Figure 10b) at both low and high confining pressure levels, except for the intermediate level ( 10 psi ). It has high positive value at granular coarse aggregates ( $0 \%$ RAP) compared with $50 \%, 75 \%$, and $100 \%$. The $75 \%$ RAP $K_{2}$ values appear to be less sound that those for the other two percentages of RAP ( $50 \%$ and $100 \%$ ).
$K_{3}$ values are almost zero for all RAP concentrations (Figure 10c) at both low and high confining pressure levels, except for the intermediate level ( 10 psi ). It has a high negative value at granular coarse aggregates ( $0 \%$ RAP) compared with those of the other percentages of RAP $(50 \%, 75 \%$, and $100 \%)$. The $75 \%$ RAP values appear to be unreasonable compared with those of $50 \%$ and $100 \%$ RAP.

The behavior of this model is approximately the same for both granular base coarse aggregates and RAP when used in a base layer. However, if the confining pressure levels are increased to 10,15 , and 20 psi , the model appears to be a better fit for the case of granular base coarse aggregates. In this case, the values of each $K$ parameter are reasonable and agree better with those shown in the literature for the MEPDG $M_{R}$ model.


FIGURE 10 Analysis of MEPDG model versus RAP concentration: confining pressure effect on (a) $K_{1}$, (b) $K_{2}$, (c) $K_{3}$, and (d) measured $M_{A}$ versus percentage of RAP.

Proving the variation of $K_{2}$ and $K_{3}$ at 10 psi for $75 \%$ RAP, Figure 10 d shows that the general trend of the original measured $M_{R}$ data for $75 \%$ RAP was lower than for both $50 \%$ and $100 \%$ RAP at all tested confining pressure levels. These results were much lower than expected for 75\% RAP as the general trend that $M_{R}$ increases when the percentage of RAP increases.

## CONCLUSIONS

Each unitless $K$ regression parameter in the MEPDG $M_{R}$ prediction model was compared for confining pressure levels and RAP concentrations at various testing field conditions, such as water content, MDD, and freeze-thaw cycles. The findings follow.
$K_{1}$ is dramatically affected by confining pressure levels for both base course layer cases: RAP and granular coarse aggregates. However, this variation is minimized at water contents close to OMC for RAP. Generally, $K_{1}$ increases with an increased percentage of RAP at confining pressure levels below 10 psi.
$K_{1}$ values at $90 \%$ MDD are higher than those at $100 \%$ MDD for all the same testing conditions for RAP and granular coarse aggregates. However, they are not significantly affected by freeze-thaw cycles under the same testing conditions.

The general trend of $K_{2}$ is that it decreases with an increasing percentage of RAP at low confining pressure levels ( $<10 \mathrm{psi}$ ) and vice versa at high confining pressure ( $>10 \mathrm{psi}$ ). Water content variation is an effective factor for $K_{2}$ values for the case of RAP. However, this effect diminishes when OMC is reached.
$K_{2}$ values increase at $90 \%$ MDD for both RAP and granular coarse aggregates. Freeze-thaw cycles do not negatively affect $K_{2}$ for both cases. Generally, $K_{2}$ values decrease at $90 \%$ MDD and freeze-thaw cycles with an increased percentage of RAP in the base layer.
$K_{2}$ values at high confining pressure levels above 10 psi are below zero, which contradicts the concept of stress stiffening related to this parameter, as confirmed by the literature survey.
$K_{3}$ values are affected by confining pressure variation, especially at intermediate levels. However, this variation is lowest at water content levels close to OMC for the RAP case.

The $K_{3}$ value is negative at low confining pressure levels ( $<10 \mathrm{psi}$ ) and water contents close to OMC. This finding appears reasonable and to satisfy the concept of shear softening related to this parameter, as confirmed in the literature. $K_{3}$ values are almost the same for the two cases of with or without RAP in the base course.

A $90 \%$ MDD is an effective factor for $K_{3}$ values for both cases of granular coarse aggregates and RAP, as the absolute values increased compared to $100 \%$ MDD at the same testing conditions. However, freeze-thaw cycles are not effective for the $K_{3}$ parameter for both cases of with or without RAP in the base layer.

The $M_{R}$ prediction by the MEPDG model fits the two studied base course cases, which are the traditional base coarse aggregates and three concentrations of a RAP-aggregate combination (50\%, $75 \%$, and $100 \%$ ) for confining pressure levels below 10 psi .

The case of $75 \%$ RAP in the base layer showed some extreme absolute values related to the trends of $50 \%$ and $100 \%$ RAP. This result is repeated in several other testing conditions.

Further research is needed to validate the MEPDG model for $M_{R}$ prediction of the RAP base layer through testing of extra parameters related to extreme environmental conditions. Also, more study is needed to relate modeling $M_{R}$ and permanent deformation for the RAP base layer.

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