

1 **Enhanced Model for Continuous Dielectric-Based Asphalt Compaction Evaluation**

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## **ABSTRACT**

The compaction of asphalt concrete significantly affects long term pavement performance. While coring provides a relatively accurate way of assessing in-place density at specific locations, the coverage of the assessment is limited, especially at longitudinal joint locations. This can be particularly problematic since it is difficult to identify problematic locations that are likely to fail prematurely using current compaction assessment methods. Ground penetrating radars (GPR) provides an attractive non-destructive testing alternative for evaluation of compaction quality, especially with recent significant improvements in the GPR technology for this specific application. However, assessment of the air void content of the asphalt mix from the GPR-measured dielectric constant of the surface requires conversion of dielectric variation to air void content variation, which is the subject of this paper. An alternative to the commonly used model is proposed, leading to more justifiable predictions for low values of dielectric constants. The proposed model was used to interpret data from a 7-mile long asphalt overlay construction project. The results of the interpretation as compared with the results obtained with the conventional model show an improvement on the stability of the prediction at low air void contents, especially when core calibration data is limited and uncertainty is considered. These results are promising in the direction of reducing field cores necessary to have a stable model providing continuous compaction assessment of new asphalt pavement construction.

*Keywords:* GPR, Compaction, Continuous, Assessment, NDT, Model

## 1 INTRODUCTION

2  
3 Past research has shown that the performance of asphalt concrete is highly dependent on the air void content  
4 of the compacted asphalt mixture. The air void content has been shown to correlate with key asphalt  
5 characteristics such as stiffness (1), strength (2) and dynamic modulus (3). Kassem, et al. (4) found that  
6 increased air void content correlated with the increased occurrence of various pavement distresses including  
7 excessive aging and moisture damage that negatively impacted long term performance. The impacts on  
8 long term performance were quantified in a study performed by Linden et al. (5), which estimated that each  
9 one percent increase in air voids over seven percent lead to an approximately ten percent reduction in  
10 pavement life.

11 Typically, asphalt compaction is assessed using coring, which is destructive, expensive, time consuming,  
12 and limited in coverage. Though these measurements are useful for post-construction analysis and are often  
13 used as primary components of quality assurance measurements, they cannot provide a real-time feedback  
14 during the paving operation. The issues associated with traditional measures of compaction create a need  
15 for nondestructive methods that can collect data continuously, cheaply, and quickly.

16 Ground penetrating radar (GPR) provides a non-destructive testing alternative that allows for walk-behind  
17 or vehicle mounted measurements (6-7). GPR uses electromagnetic waves to explore subsurface  
18 characteristics. In transportation infrastructure survey, GPR has been commonly applied to detect free  
19 water (8), to estimate the dielectric property of pavement materials (9), to estimate the layer thicknesses  
20 (10), and asphalt concrete layer density (11-14). ASTM standard ASTM D6432-11 provides a procedure of  
21 applying GPR for subsurface investigation.

22  
23 Determination of dielectric properties of the asphalt layer with GPR has been traditionally done through  
24 measurement either round trip travel time to reflection at the depth of the AC layer or surface reflection.  
25 The travel time approach covers a greater depth, but relies on a known thickness. The asphalt thickness is  
26 often unknown and spatially variable. Moreover, if the asphalt layer is placed in several lifts or as an  
27 overlay over an existing asphalt pavement, it may be difficult to separate the travel time in the individual  
28 lifts from the overall travel time of the electromagnetic signal in the asphalt layer.

29  
30 The AC surface reflection method uses the ratio of the amplitude of the GPR signal reflection from air to  
31 the asphalt surface,  $A_o$ , to the incident amplitude (represented by the reflection from the metal plate),  $A_i$ , to  
32 determine the bulk dielectric constant of the asphalt,  $e_r$ . The dielectric constant of the surface is determined  
33 according to Saarenketo and Scullion (15) using the following equation:

$$34 \quad e_r = \left( \frac{1 + \left(\frac{A_o}{A_i}\right)}{1 - \left(\frac{A_o}{A_i}\right)} \right)^2 \quad (1)$$

35 The advantage of this approach is that if the upper lift is sufficiently thick (thicker than 30 mm) then the  
36 measured AC surface reflection depends only on the properties of the upper layer.

37 For newly placed asphalt lift, the dielectric constant values determined from Equation 1 can be empirically  
38 related to the relative ratio of pore volume to the total volume of each specific asphalt mix since air has a  
39 lower dielectric constant than the surrounding asphalt material and the aggregate type and volumetric  
40 proportion are typically uniform (16-17). Since dielectric properties of the asphalt mix depend on the  
41 dielectric properties of other components of the mix which vary from project to project a universal dielectric  
42 constant to air void content conversion is not feasible. To account for these changes, cores need to be taken  
43 for each new mix at locations where conversion from dielectric constant to air void content is conducted.  
44 The correlation between the air voids and dielectric constants plays a key role in the accuracy of the air

1 voids predictions made using the model.

2 There have been several field trials of GPR for nondestructive testing (NDT) determination of air voids.  
3 The first large scale trial was performed in Finland in 1996-1997 (18). Recently, several state DOTs in the  
4 USA have held trials of the technology (19-22). The most notable recent application of GPR for compaction  
5 surveying was conducted by Sebesta, Scullion and Saarenketo (16) as part of the SHRP2 RO6C activities.  
6 Though several implementations have consistently demonstrated the ability of GPR technology to provide  
7 real-time information on relative compaction, use of GPR to estimate actual air void content has limitations  
8 and is still a challenging problem in spite of the significant improvements in the GPR technology for this  
9 specific application under the SHRP-2 project (16).

10 Various impulse radar versions of ground penetrating radar have shown that the dielectric properties  
11 determined from the asphalt surface reflection amplitude corresponds with core measured air void content  
12 (18,22, 23). Additionally, a step frequency array-based method improves the coverage and productivity of  
13 the measurements, making it an attractive alternative to current state-of-the-practice procedures (21). While  
14 these studies showed the potential of new technology for improved quality assurance in selected locations,  
15 the focus of this study is on how a stable compaction assessment can be achieved in full-scale  
16 implementation. In the case of the step-frequency array system (24), these technologies can require  
17 intensive data processing from the frequency domain or can be cost prohibitive, while the single impulse  
18 array systems do not provide necessary coverage for widespread implementation.

19  
20 The GPR equipment used in this study, the rolling density meter (RDM), is based on a system that evolved  
21 from recent research conducted under a National Academies of Science sponsored Strategic Highway  
22 Research Program (SHRP-2) (16). It uses similar antenna, but also applied in a three-channel array to  
23 obtain some of the benefits in coverage explained in (21), where multiple antenna pairs are used in each  
24 pass.

25  
26 This paper deals with a problem of conversion of dielectric variation to air void content variation.  
27 Traditionally, a simple exponential model is used for dielectric – air void content correlation. The current  
28 practice is to develop this model for each construction project to minimize the error caused by variations in  
29 mix design and properties. To minimize extrapolation outside the calibration limits, it is recommended to  
30 collect cores representing full range survey dielectric constant values (18), (19), (21). However, this is not  
31 always practical since the full range of dielectrics on a project are not known until the project is complete.  
32 Moreover, areas with a high air void content often exhibit higher local variation causing higher uncertainty  
33 in the core air void measurements. Since only a limited number of cores with a high air void content is  
34 available it increases the uncertainty in the dielectric property – air void relationship for high air void  
35 content values. At the same time, an accurate conversion of the dielectric values to the air void content is  
36 especially important for the areas with low measured dielectric values, because they often determine  
37 whether the pavement will fail prematurely.

38  
39 The paper presents an approach for development of a modified model to convert dielectric variation to air  
40 void content variation. The paper considers electromagnetic mix modeling theory model and is based on  
41 the conventionally accepted empirical method. The proposed model includes physics-based constraints to  
42 ensure more reliable determination of the areas with high air void content.

## 43 44 **OVERVIEW OF ELECTROMAGNETIC MIX MODELING APPROACH**

45  
46 Several approaches have been developed for relating the dielectric properties of an asphalt mix to its air  
47 void content. The most commonly used approach is the use of an empirical correlation between these two  
48 mix characteristics. Many studies successfully used the exponential model (Conventional Model) (15, 16,  
49 21, 22):  
50

$$AV = A * \exp(-B * \varepsilon) \quad \text{Eq. 1}$$

where, AV is the air void content,  $\varepsilon$  is the input hot mix asphalt dielectric constant, A and B are calibration constants.

While this purely empirical relationship is effective in converting dielectric constant to air void content when necessary dielectric precision is achieved and sufficient core calibrations are available, a more rigorous approach should derive the air void- dielectric constant relationship for the asphalt mix from the dielectric properties and volume fractions of the asphalt mix components. While this ideal is not fully achievable using the currently available mix models, the mix-modeling approach should be considered to ensure a more stable physics-based empirical model.

Electromagnetic mix-modeling has been developed for dielectric characterization of rock properties such as estimation of rock porosity (25), water and clay influences (26), and other in geophysical applications (27). This work was further extended for application to civil engineering materials including evaluation of hot mix asphalt using ground penetrating radar (23, 28-30). The mix model method is based on the modelling of how an electromagnetic wave interacts with composite materials using assumptions for how the wave interacts with the different asphalt concrete components. The mix models evaluated in this study differ only in form, as the coefficients used (dielectric and volumetric values) are the same in each model. A more detailed description of each model can be found elsewhere (23, 28-30). Equations 2 through 4 show commonly accepted mix models for asphalt concrete.

$$\text{Rayleigh Model} \quad AV = 1 - \frac{\frac{\varepsilon_{HMA} - \varepsilon_b}{\varepsilon_{HMA} + 2 * \varepsilon_b} \frac{1 - \varepsilon_b}{1 + 2 * \varepsilon_b}}{G_{mm} * \left( \frac{\varepsilon_s - \varepsilon_b}{\varepsilon_s + 2 * \varepsilon_b} \right) * \left( \frac{1 - P_b}{G_{sb}} \right) - \frac{(1 - \varepsilon_s)}{1 + 2 * \varepsilon_b} * \left( \frac{1}{G_{mm}} \right)} \quad \text{Eq. 2}$$

$$\text{Bottcher Model} \quad AV = 1 - \frac{\frac{\varepsilon_{HMA} - \varepsilon_b}{3 * \varepsilon_{HMA}} \frac{1 - \varepsilon_b}{1 + 2 * \varepsilon_{HMA}}}{G_{mm} * \left( \frac{\varepsilon_s - \varepsilon_b}{\varepsilon_s + 2 * \varepsilon_{HMA}} \right) * \left( \frac{1 - P_b}{G_{sb}} \right) - \frac{(1 - \varepsilon_b)}{1 + 2 * \varepsilon_{HMA}} * \left( \frac{1}{G_{mm}} \right)} \quad \text{Eq. 3}$$

$$\text{CRIM for } \alpha=2 \quad AV = 1 - \frac{\sqrt{\varepsilon_{HMA}} - 1}{G_{mm} * \left( \frac{P_b}{G_b} \sqrt{\varepsilon_b} + \frac{(1 - P_b)}{G_{sb}} \sqrt{\varepsilon_s} - \frac{1}{G_{mm}} \right)} \quad \text{Eq. 4}$$

where, AV is the air void content,  $\varepsilon_{HMA}$  is the input hot mix asphalt dielectric constant, and the remaining variables are given in Table 1.

The mixing models are advantageous as compared to a purely empirical approach in that they are rational. These models – combined with laboratory testing and more accurate assumptions – may eventually reduce or even eliminate the need for calibration cores. These models guarantee reasonable trends, i.e. sensitivity to the properties of the individual components, so it may make them less susceptible to errors. For example, each model can be verified for the extreme conditions, such as predicting 100% air void content (0 bulk specific gravity) when a dielectric constant of 1 is used. By comparison, the conventional purely empirical model does not hold true outside of the measured dielectric ranges.

Al-Qadi et al. conducted a sensitivity study and evaluated the CRIM, Rayleigh, and Bottcher models for characterizing asphalt compaction (23). A method of calibrating the models using

1 laboratory mixed and compacted specimens (23). Table 1 gives the assumed values used in the  
2 sensitivity study as well as the values obtained from a least squares regression matching the  
3 laboratory measured air voids.

4  
5 Figure 1 shows the air void content measured and predicted versus measured dielectric using the  
6 assumptions reported by Al-Qadi et al (23). The plot is adapted to convert the presented bulk  
7 specific gravity relationship to air void content which is compatible with the results of this study  
8 by using the assumed max specific gravity reported of 2.526. The results from the Al-Qadi study  
9 show a reasonable agreement between the measured and predicted values, although the measured  
10 data is less sensitive to changes in dielectric than the predicted response. Additionally, it should  
11 be noted that the air void contents were greater than typical air voids observed in the field due to  
12 limitations in their compaction process when preparing laboratory samples.

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15  
16 While the Al-Qadi et al. showed discrepancies and limitations with the laboratory calibration  
17 method and ability to match the sensitivity of measured air void content versus dielectric values,  
18 the Rayleigh model was determined to be the most rational mix model of the available choices  
19 (23). Although imperfect, this model can be used to quantify the expected effect of mix changes.  
20 For example, Figure 2 shows the Rayleigh model predicted air void content versus dielectric  
21 constant using the assumed values from Table 1 to compare the effect of aggregate dielectric and  
22 binder dielectric changes. The aggregate dielectric constant (top) shows 3 aggregate value plots  
23 that represent a reasonable Minnesota source granite  $e=5$  (green), Al-Qadi et al. assumed limestone  
24 aggregate ( $e=6$ ), and reasonable Minnesota source limestone aggregate ( $e=7$ ). These represent a  
25 relatively conservative range as compared to the 4 to 9 dielectric values of aggregates reported by  
26 multiple studies (23, 27). The binder dielectric content (bottom) of 2, 3, and 4 represent a relatively  
27 large range of dielectric values compared to the 2.5 and 3.0 values used by Araujo et al. and Al-  
28 Qadi et al., respectively (23,27). It can be observed that a significantly greater effect on air void  
29 content is caused by changes in the aggregate dielectric as compared to the binder, even for  
30 relatively small changes in possible aggregate sources. For example, Araujo reported dielectric  
31 constants from 24 samples ranged from 4 to 9 including various sedimentary, magmatic, and  
32 metamorphic rocks, which matched previous studies by Ulaby et al. (31) and Parkhomenko (32).  
33 The significance of aggregate dielectric constant is a reasonable result considering the mixing rules  
34 are based on volumetric and assumed or calculated dielectric constant values of each component,  
35 and the aggregate contributes the highest percentage of the asphalt pavement volume. These types  
36 of observations are valuable in determining when the type and magnitude of mix changes may not  
37 require unique calibration and mitigate the need for additional field cores.

38  
39 The significant effect of aggregate dielectric on measured dielectric has been recognized and a  
40 method of measuring the aggregate dielectric using a portable network analyzer and two  
41 cylindrical cavities to measure the dielectric constant of various aggregate sources has been  
42 developed (27, 29, 30). While this approach is especially attractive in the direction of eventually  
43 eliminating the need for field cores, this method reported precision issues involving difficulties  
44 with creating a rock specimen at adequate dimensions, requires significant user expertise to run  
45 the test, and has to be conducted at the same frequency content as the GPR equipment used in the  
46 field. The Al-Qadi et al. and Araujo et al. studies included the Rayleigh model (least sensitive  
47 model) as the most effective model representation of asphalt pavement (23, 27), which is used for

1 comparison with the method proposed in this paper.  
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## 6 **OBSERVED EMPIRICAL RELATIONSHIPS VERSUS MIX MODEL PREDICTIONS** 7

8 Both mix model theory and observed field relationships show that the measured dielectric is a function of  
9 the type of pavement being tested in addition to the well-established relationship between air void content  
10 and dielectric constant for a given mix. Therefore, while the rolling density meter method for assessment  
11 of compaction is valid when the pavement material is consistent, comparison of different pavement designs  
12 within a project may not be appropriate. For example, a pavement with a higher dielectric aggregate source  
13 like limestone may have a higher overall dielectric profile than a pavement with a granite aggregate source  
14 that has a low dielectric value regardless of air void content. Figure 3 gives an example of core measured  
15 air void content versus RDM measured dielectric constants on a pavement test-track with 6 different  
16 pavement mix designs at MnROAD. These measurements were made without any assumptions about mix  
17 components. It can be observed that a relatively good agreement between core and RDM measurement  
18 dielectrics is obtained even when combining data from 5 different mix designs as long as the aggregate  
19 source was constant (granite in this case). It can also be observed that the limestone-based mix exhibited  
20 higher dielectric values than the granite-based mix regardless of the air void content. This agrees with the  
21 mix-model predictions shown in figure 2, where changes other than aggregate dielectric constant are  
22 relatively small in comparison to the change in dielectric caused by changes in air void content.  
23

24  
25 While the empirical data generally agrees with the predicted mix model relationships, the observed  
26 sensitivity (change in air voids for a given change in dielectric) show a similar discrepancy to that observed  
27 in the laboratory study conducted by Al-Qadi et al., even when using the Rayleigh mix-model (23). Figure  
28 4 shows curves developed from multiple locations (Texas, Minnesota, Maine, Nebraska), conducted by  
29 multiple agencies (University of Minnesota, Minnesota Department of Transportation, and Texas  
30 Transportation Institute). It can be observed that each model shows a less sensitive slope between  
31 measured and mix model predicted air void content as compared to the curves shown in figure 2. A  
32 superposition of an equivalent mix-model curve from figure 2 (red) along with embedded arrows in each  
33 figure placed at the granite aggregate predicted and observed curves, it can be observed that the same change  
34 from 9% air void content to 5% air void content requires over twice as much of a change in observed  
35 dielectric as the Rayleigh model predicts. This agrees with the observation from the laboratory study  
36 showing the mix-models to be overly sensitive to changes in dielectric (23).  
37

## 38 **PROPOSED MODEL** 39

40 The practicality and sensitivity issues described in the previous section preclude direct  
41 implementation of the mixing models. However, the value of incorporating a rational model in  
42 the conversion is proposed to address issues with the currently accepted conventional model.  
43 The currently accepted conventional model can be unstable in predicting air void content at the  
44 extremes (it is recommended to collect 5<sup>th</sup> and 95<sup>th</sup> percentile dielectric values to avoid  
45 extrapolation) (18, 19, 21), especially if there is a lack of dielectric versus core data to get an  
46 accurate representation. This is known from previous studies and it is recommended to select  
47 field core locations from a wide range of dielectric data to calibrate when using the currently  
48 accepted conventional method (18, 19, 21). However, selecting a wide range of core locations is  
49 not always practical. Moreover, due to a small number of replicates at the end of the measured  
50 ranges, a small error in the measured either dielectric value at the core location or air void value

1 of the core may lead to a significant error in prediction, especially outside of the calibration  
 2 range. This instability is magnified when extrapolating outside of the core calibrated range when  
 3 using a model that is not constrained.

4  
 5 To evaluate the instability of the current model, consider data collected with a RDM during  
 6 construction of 7-miles of a 1.5-in asphalt overlay of TH14 near Eyota, MN. Figure 5 shows 32  
 7 core measured air void results versus RDM measured dielectrics along with the conventional  
 8 model shown in bold. It can be observed that the conventional model matches core data well.  
 9 However, the conventional model predictions become unrealistic for lower dielectric values.  
 10 When the dielectric constant approaches 1 a predicted air void percentage approaches 337.3%.

11  
 12  
 13 To address this limitation, the following model is proposed:  
 14  
 15

$$16 \quad AV = \exp\left(-B\left(D\left|\frac{1}{e-C}-\frac{1}{1-C}\right| - 1\right)\right)$$

17  
 18 where AV is the air void content, e is an input dielectric constant, B, C, and D are constants  
 19 determined by a non-linear least-squared fit.

20  
 21 The proposed approach is an empirical model allowing for a slope that matches the observed field  
 22 data and is constrained to physical behavior even at low dielectric/high air void content locations.  
 23 If the dielectric constant input approaches 1 then the model predicts the air void content to be  
 24 100%, which is the correct value. This constraint makes the proposed model less sensitive to errors  
 25 in the calibration data for cores with high air void content and more suitable for extrapolation  
 26 beyond the lowest dielectric value in the calibration dataset.

27  
 28 The proposed model does not offer a significant improvement over the currently accepted model  
 29 when many cores are available and the range of core calibrations span the range of collected  
 30 dielectric values for interpolation between known points, such as was the case in the field collected  
 31 data in this study. However, the desire to limit cores, logistical challenges such as accommodating  
 32 moving traffic closures, and other factors may not always be conducive to gathering a significant  
 33 amount of cores ranging the full measured dielectric range. These cases, where extrapolation  
 34 beyond the calibrated range with limited cores show the value of the proposed mechanistic-based  
 35 model.

36  
 37 To demonstrate the advantage of the proposed model, both conventional and the proposed models  
 38 were re-calibrated using the same core data except where the measured dielectric was 5.3 or lower.  
 39 Removing these highest dielectric values still provided a reasonable amount of calibration data  
 40 with 12 remaining cores. To illustrate the stability of the proposed model in comparison to the  
 41 currently accepted model, figure 6 shows the predicted air void content versus a large range of  
 42 dielectric values for both the conventional (solid black line) and proposed (dashed black line)  
 43 models. Figure 6a shows the predicted air void contents with the full range of data. It can be  
 44 observed from Figure 6 that within the calibration range, i.e. interpolated dielectric to air void  
 45 conversions, both the conventional model and the proposed model predict similar air void contents.  
 46 However the conventional model predicts a significantly higher air void content at low dielectric



1 ranges. Comparison with figure 5, where the full dataset was available, shows that the calibrations  
2 with the modified dataset resulted in much greater discrepancy between the conventional and  
3 proposed model predictions than the calibrations with the original dataset. To put the discrepancy  
4 in perspective of realistically measured dielectrics for the project, figure 6b shows the same  
5 comparison within a small range of dielectrics spanning the confined joint measured dielectric  
6 distribution (from 4.75 to 5.25).

7  
8 It can be observed that significant discrepancies occur, even within realistic dielectric ranges with  
9 the predicted air void content from the conventional model at the low end of 4.75 resulting in a  
10 17% air void content compared to less than 12% using the proposed method. Comparison with  
11 the full dataset predictions, expected realistic air void content, and measured core air void content  
12 support the proposed model prediction as compared to the conventional model. These observations  
13 lead to the conclusion that the proposed model offers a significant improvement of the predictions  
14 for the low dielectric value areas, especially when the accuracy of the measurements are low or an  
15 insufficient number of cores on the extremes (high or low dielectric values) is available.

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### FIELD TESTING RESULTS

As observed above, the conventional and modified model produce very similar results for a wide range of the measured dielectric values, except when the dielectric values extend to low dielectric and air void contents. While the conventional model leads to unreliable and physically inadmissible air void contents for low dielectric values, the modified model offers a more rational and justifiable alternative. To evaluate the significance of this enhancement when applied to a full-scale field implementation, consider the data collected with a RDM from 7 miles of a 1.5-in asphalt overlay of TH14 near Eyota, MN. Figure 7 shows histograms of dielectric values measured with the RDM along a stretch of pavement and separated into 2 groups:

- Mat – a GPR sensor path is at least 2 ft from the closest joint
- Confined Joint – a GPR sensor path is within 1 ft of the longitudinal joint compacted when the adjacent lane has been already constructed.

Using the conventional and modified calibration models, dielectric frequency distributions were converted into the distributions of the relative densities, RD, defined as

$$RD = 1 - AV$$

Figure 8a shows the histograms obtained from these distributions with the conventional model while Figure 8b shows the histograms obtained with the proposed model. Table 2 presents a summary statistics obtained using these distributions. Comparisons of the Figures 8a and 8b reveals that the relative density distributions for the mat are almost identical. Table 2 shows that these distributions have the same mean and median values. The relative densities corresponding to 95% percent within limits are also very similar: 91.0 and 91.2 percent from the conventional and proposed models, respectively. The confined joint exhibited a lower relative density compared to the mat compaction level, as expected. It is also not a surprise that a greater discrepancy between the relative density distributions is observed for the confined joint. Although the mean and median values are also very similar, the difference between the relative densities corresponding to 95% percent within limits is significantly greater: 89.0 and 89.5 percent from the conventional and proposed models, respectively. This suggests that the conventional model tends to overestimate the percentage of the area with low compaction. Similar analysis of the unconfined side of the joint showed an even greater difference between the models.

### CONCLUSIONS

Early deterioration and long term performance of asphalt pavements is highly affected by quality of compaction. To minimize potential delay in traffic closure on rehabilitation of heavily trafficked areas where quality of compaction is especially important, it is often desirable that the measurements can be taken over longer sections in a short timeframe immediately after final roller compaction, while still providing the necessary pavement coverage. An accurate conversion of the dielectric values to the air void content is especially important for the areas with low measured dielectric values such as longitudinal joints, since they often cause early pavement failure.

This study shows the potential of compaction assessment using a SHRP2 recommended technology referred to as the rolling density meter (RDM), which assesses compaction using a

1 continuous dielectric profile. A key to the success of this continuous, non-destructive technology  
2 is relating the dielectric constant to actual achieved air void content. Several approaches for  
3 development of an implementable model to convert dielectric variation to air void content variation  
4 were considered in the paper including the conventional empirical model and physics-based  
5 electromagnetic mixing models. While the empirical model often produces reliable predictions,  
6 they require a wide range of calibration cores to ensure stability, and produce unrealistic results  
7 for low dielectric constant values. The mixing models are advantageous in that they are derived  
8 from the dielectric properties and volume fractions of the asphalt mix components. However, it  
9 was found that the currently available mix models do not correctly predict the change in the air  
10 voids at a similar sensitivity to those observed in the field by multiple studies. A modified  
11 empirical model is introduced, matching the accuracy of prediction within the core calibration  
12 range, and improving the prediction at low air void contents. The improvement is magnified when  
13 a lack of core data and uncertainty is considered at low air void contents. The findings when using  
14 the modified empirical model as applied to a full scale construction project are promising due to  
15 the implications in reduction of field cores necessary to convert dielectric measurements to a  
16 continuous compaction assessment of new asphalt pavement construction.

## 19 REFERENCES

- 22 1. Bonnaure, F., et al. "New method of predicting the stiffness of asphalt paving  
23 mixtures." *Association of Asphalt Paving Technologists Proc.* Vol. 46. 1977.
- 24 2. Pellinen, T. K., Jiansheng Song, and Shangzhi Xiao. "Characterization of hot mix asphalt  
25 with varying air voids content using triaxial shear strength test." *Proceedings of the 8th*  
26 *Conference on Asphalt Pavements for Southern Africa (CAPSA'04)*. Vol. 12. 2004.
- 27 3. Witczak, M., and O. Fonseca. "Revised predictive model for dynamic (complex) modulus  
28 of asphalt mixtures." *Transportation Research Record: Journal of the Transportation*  
29 *Research Board* 1540 (1996): 15-23.
- 30 4. Kassem, E., et al. "Measurements of asphalt pavement density using ground penetrating  
31 radar and its relationship to performance." *Transportation Research Board 91st Annual*  
32 *Meeting*. No. 12-4051. 2012.
- 33 5. Linden, R. N., Joe P. Mahoney, and Newton C. Jackson. "Effect of compaction on  
34 asphalt concrete performance." *Transportation Research Record* 1217 (1989).
- 35 6. Evans, R., M. Frost, M. Stonecliffe-Jones, and N. Dixon. "A review of pavement  
36 assessment using Ground Penetrating Radar (GPR)," *Proceedings of the 12th*  
37 *international conference on Ground Penetrating Radar*, Birmingham, UK (2008)
- 38 7. Lai, W., T. Kind, S. Kruschwitz, J. Wöstmann, and H. Wiggenhauser. "Spectral  
39 absorption of spatial and temporal ground penetrating radar signals by water in  
40 construction materials," *Nondestructive Testing Evaluation International*, Vol. 67, pp. 55-  
41 63 (2014).
- 42 8. Al-Qadi, I/ L., et al. "Effect of moisture on asphaltic concrete at microwave  
43 frequencies." *IEEE Transactions on Geoscience and Remote sensing* 29.5 (1991): 710-  
44 717.
- 45 9. Al-Qadi, I. L., S. Lahouar, and A. Loulizi. "In situ measurements of hot-mix asphalt  
46 dielectric properties." *NDT & e International* 34.6 (2001): 427-434.

- 1 10. Al-Qadi, I. L., and S. Lahouar. "Measuring layer thicknesses with GPR—Theory to  
2 practice." *Construction and building materials* 19.10 (2005): 763-772.
- 3 11. Leng, Z. (2011) Prediction of In-Situ Asphalt Mixture Density Using Ground Penetrating  
4 Radar: The Theoretical Development and Field Verification, Ph.D. Thesis, University of  
5 Illinois at Urbana-Champaign, USA.
- 6 12. Leng, Z., I. L. Al-Qadi, and Samer Lahouar. "Development and validation for in situ  
7 asphalt mixture density prediction models." *NDT & e International* 44.4 (2011): 369-375.
- 8 13. Shangguan, P., and I. L. Al-Qadi. "Content-based image retrieval approaches to interpret  
9 ground penetrating radar data." *Construction and Building Materials* 69 (2014): 10-17.
- 10 14. Shangguan, P., and I. L. Al-Qadi. "Calibration of fdtd simulation of gpr signal for asphalt  
11 pavement compaction monitoring." *IEEE Transactions on Geoscience and Remote*  
12 *Sensing* 53.3 (2015): 1538-1548.
- 13 15. Saarenketo, T., and T. Scullion. "Road evaluation with ground penetrating  
14 radar." *Journal of applied geophysics* 43.2 (2000): 119-138.
- 15 16. Sebesta, S., T. Scullion, and T. Saarenketo. *Using infrared and high-speed ground-*  
16 *penetrating radar for uniformity measurements on new HMA layers*. Transportation  
17 Research Board, 2013.
- 18 17. Leng, Z., and I.L. Al-Qadi. "An innovative method for measuring pavement dielectric  
19 constant using the extended CMP method with two air-coupled GPR systems." *NDT & E*  
20 *International* 66 (2014): 90-98.
- 21 18. Saarenketo, T., and P. Roimela. "Ground penetrating radar technique in asphalt pavement  
22 density quality control." *Proceedings of the seventh international conference on ground*  
23 *penetrating radar*. Vol. 2. 1998.
- 24 19. Popik, M., K. Maser, and C. Holzschuher. "Using high-speed ground penetrating radar for  
25 evaluation of asphalt density measurements." *annual conference & exhibition of the*  
26 *transportation association of Canada, Canada*. 2010.
- 27 20. Wilson, B. T., and S, Sebesta. "Comparison of Density Tests for Thin Hot-Mix Asphalt  
28 Overlays." *Transportation Research Record: Journal of the Transportation Research*  
29 *Board* 2504 (2015): 148-156.
- 30 21. Hoegh, K., Khazanovich, L., Dai, S., and H.T. Yu "Evaluating asphalt concrete air void  
31 variation via GPR antenna array data." *Case Studies in Nondestructive Testing and Evaluation*3  
32 (2015): 27-33.
- 33 22. Maser, K., and A. Carmichael. *Ground Penetrating Radar Evaluation of New Pavement*  
34 *Density*. No. WA-RD 839.1. 2015.
- 35 23. Al-Qadi, I., Z. Leng, S. Lahouar, and J. Baek. "In-place hot-mix asphalt density  
36 estimation using ground-penetrating radar." *Transportation Research Record: Journal of*  
37 *the Transportation Research Board* 2152 (2010): 19-27.
- 38 24. Scott, M. L., N. Gagarin, M.K. Mills, and M. Oskard. "Step Frequency Ground  
39 Penetrating Radar Applications to Highway Infrastructure Measurement and System  
40 Integration Feasibility with Complementary Sensors." AIP Conference Proceedings.  
41 (2006): 1, 1624-1631
- 42 25. Tong, Ma., and H. Tao. "Permeability estimating from complex resistivity measurement  
43 of shaly sand reservoir." *Geophysical Journal International* 173.2 (2008): 733-739.

- 1 26. Bona, N., A. Ortenzi, and S. Capaccioli. "Advances in understanding the relationship  
2 between rock wettability and high-frequency dielectric response." *Journal of Petroleum*  
3 *Science and Engineering* 33.1 (2002): 87-99.
- 4 27. Araujo, St., et al. "Rock permittivity characterization and application of electromagnetic  
5 mixing models for density/compactness assessment of HMA by means of step-frequency  
6 radar." *Near Surface Geophysics* 14.6 (2016): 551-562.
- 7 28. Benedetto, A., and F. Tosti. "Inferring bearing ratio of unbound materials from dielectric  
8 properties using GPR: the case of Runaway Safety Areas." *Airfield and Highway*  
9 *Pavement 2013: Sustainable and Efficient Pavements*. 2013. 1336-1347.
- 10 29. Fauchard, C., B. Beaucamp, and Laurent Laguerre. "Non-destructive assessment of hot  
11 mix asphalt compaction/density with a step-frequency radar: case study on a newly paved  
12 road." *Near Surface Geophysics* 13.3 (2015): 289-297.
- 13 30. Fauchard, C., et al. "Determination of the compaction of hot mix asphalt using high-  
14 frequency electromagnetic methods." *NDT & E International* 60 (2013): 40-51.
- 15 31. Ulaby, F. T., et al. "Microwave dielectric properties of dry rocks." *IEEE Transactions on*  
16 *Geoscience and Remote Sensing* 28.3 (1990): 325-336.
- 17 32. Parkhomenko, E. I. *Electrical properties of rocks*. Springer Science & Business Media,  
18 2012.

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1 **Tables**

2

3 **Table 1. Mix Model Inputs (23).**

Model		Aggregate ( $\epsilon_s$ )	Binder ( $\epsilon_b$ )	Maximum Bulk Spec Grav ( $G_{mm}$ )	Spec Grav of Binder ( $G_b$ )	Binder Content ( $P_b$ )	Bulk SG of Aggregate ( $G_{sb}$ )
Rayleigh	Assumed	6	3	2.521	1.015	5%	2.705
	Regression	7.44	2.00	2.276	1.015	4.1%	2.610
Bottcher	Assumed	6	3	2.521	1.015	5%	2.705
	Regression	8.33	7	3.026	1.015	6.0%	2.610
Crim	Assumed	6	3	2.521	1.015	5%	2.705
	Regression	6.38	3.21	2.602	1.015	5.1%	2.610

4

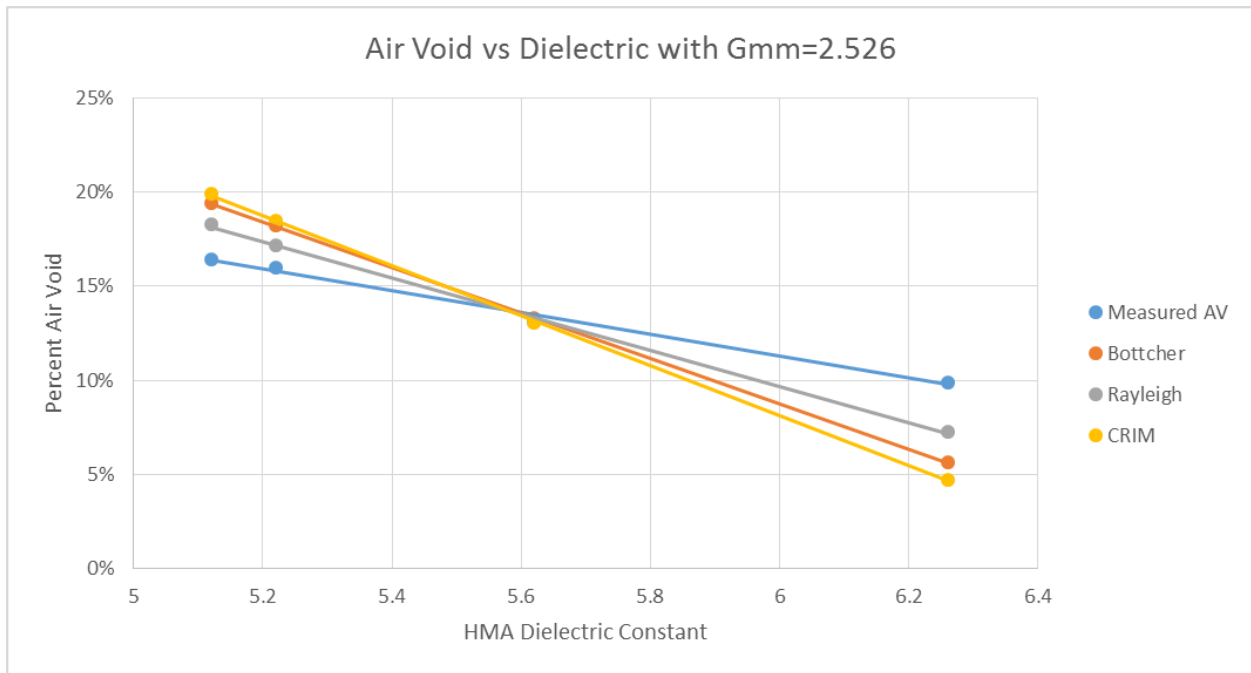
1 Table 2. Summary statistics for mat and confined joint RDM scanned locations.

	<b>Relative Density</b>			
	Conventional Model		Proposed Model	
	Mat	Confined Joint	Mat	Confined Joint
<b>Mean</b>	92.9%	91.5%	92.9%	91.5%
<b>Median</b>	92.9%	91.4%	92.9%	91.5%
<b>Low compaction end of 95% percent within limits</b>	91.0%	89.0%	91.0%	89.5%

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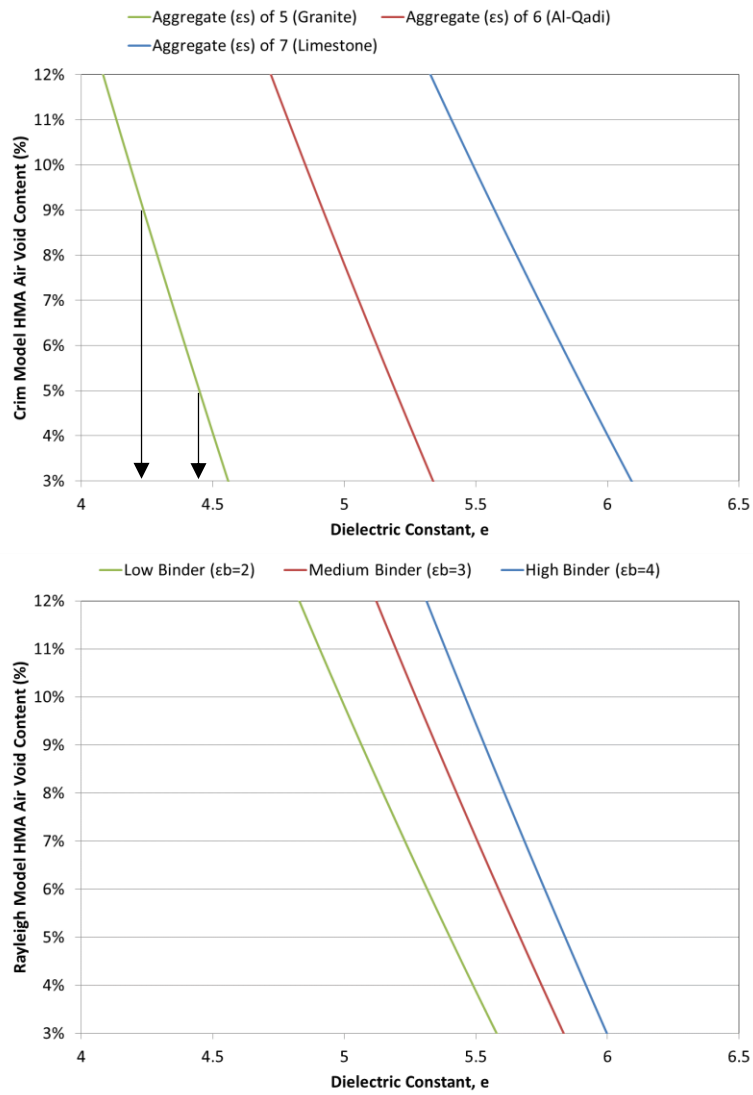
1 **Figures**  
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3  
4 Figure 1. Measured (blue) versus predicted air void content using the CRIM, Rayleigh, and  
5 Bottcher models (23).  
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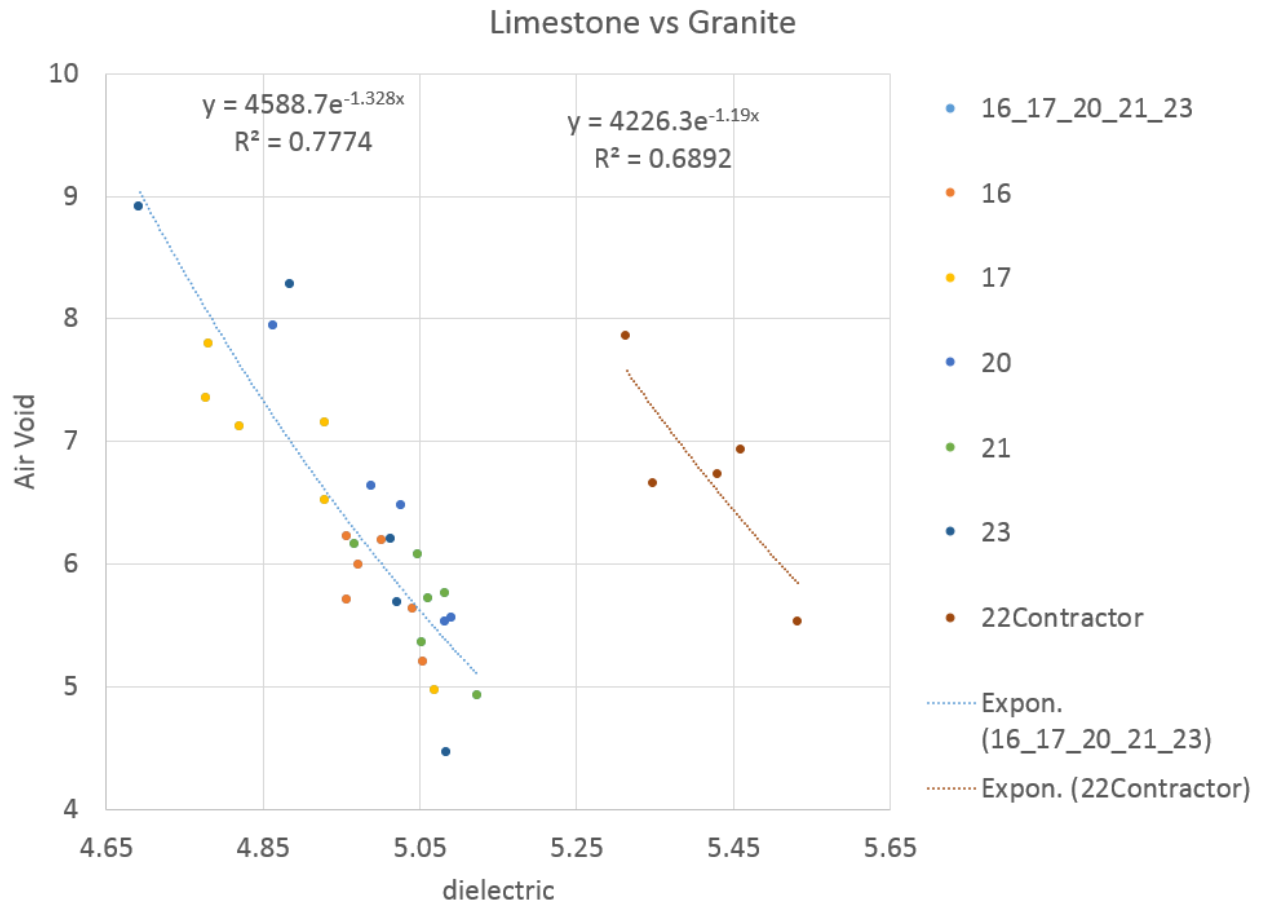


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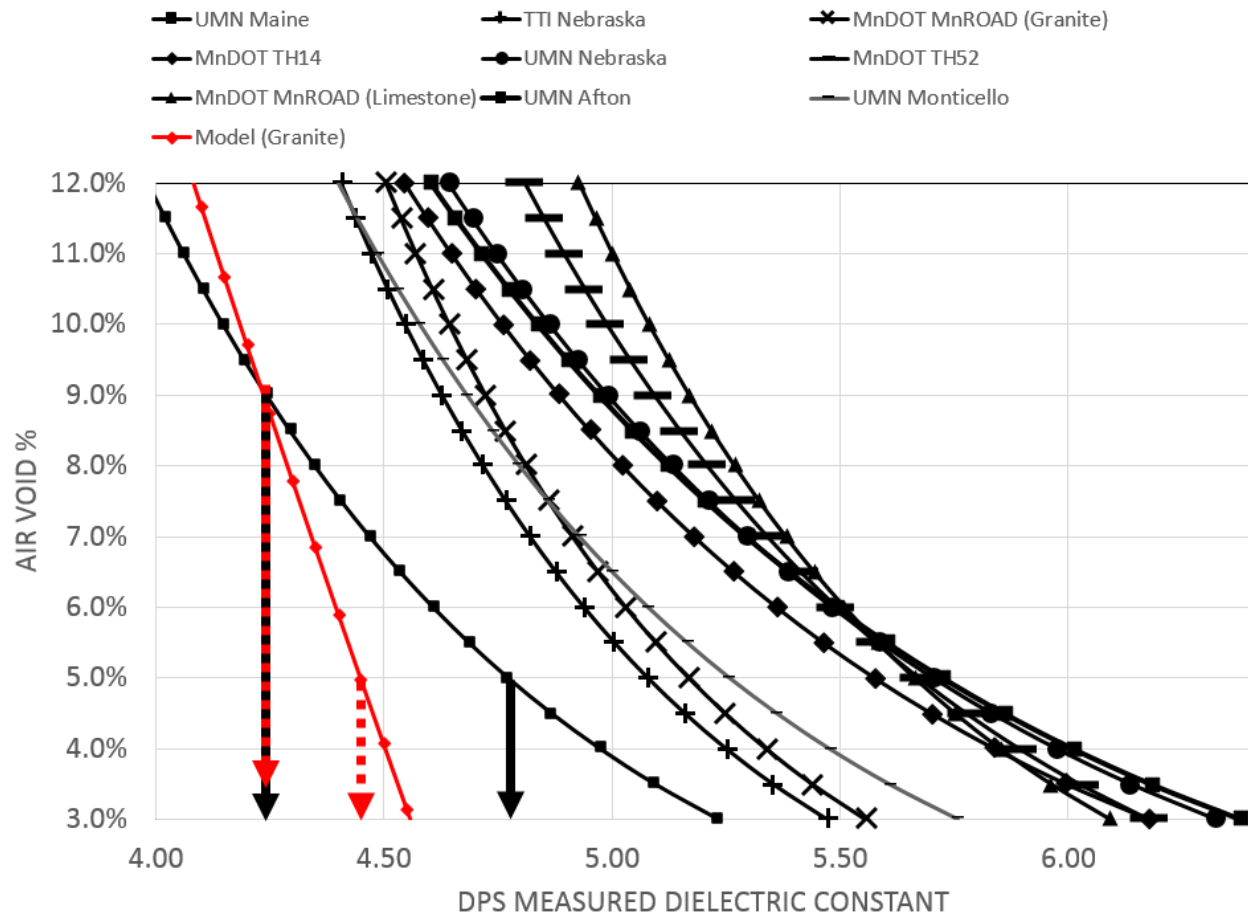
2  
3 Figure 2. Predicted change in air void content using the Rayleigh model for a range in dielectric  
4 values assumed for aggregate (top) and binder (bottom).  
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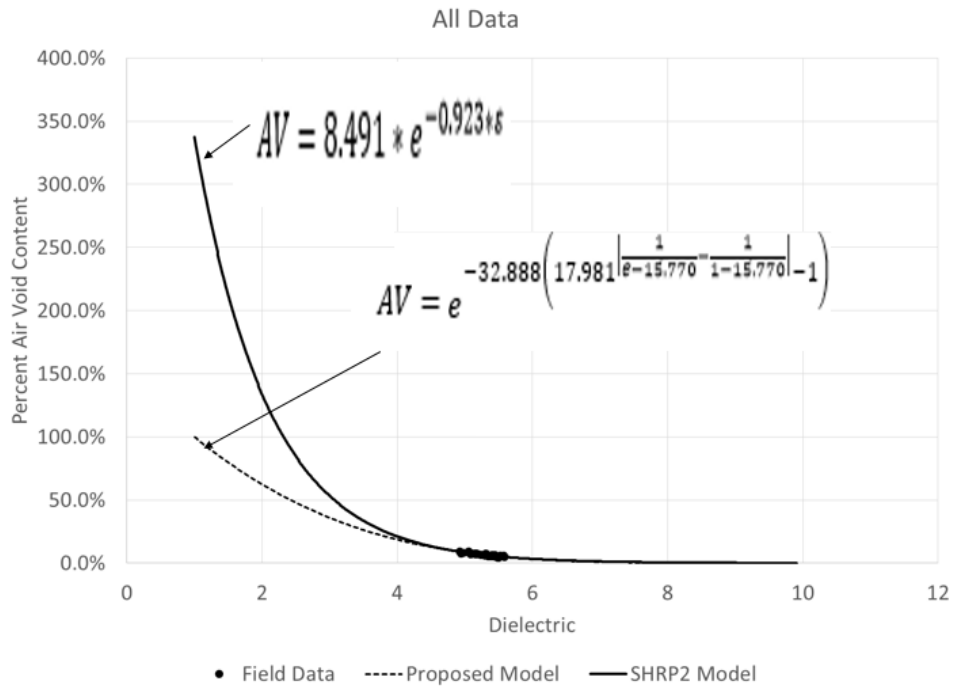


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Figure 3 Example core versus dielectric data illustrating when different mixes can be combined (left curve) and when they need to be separated (right curve).

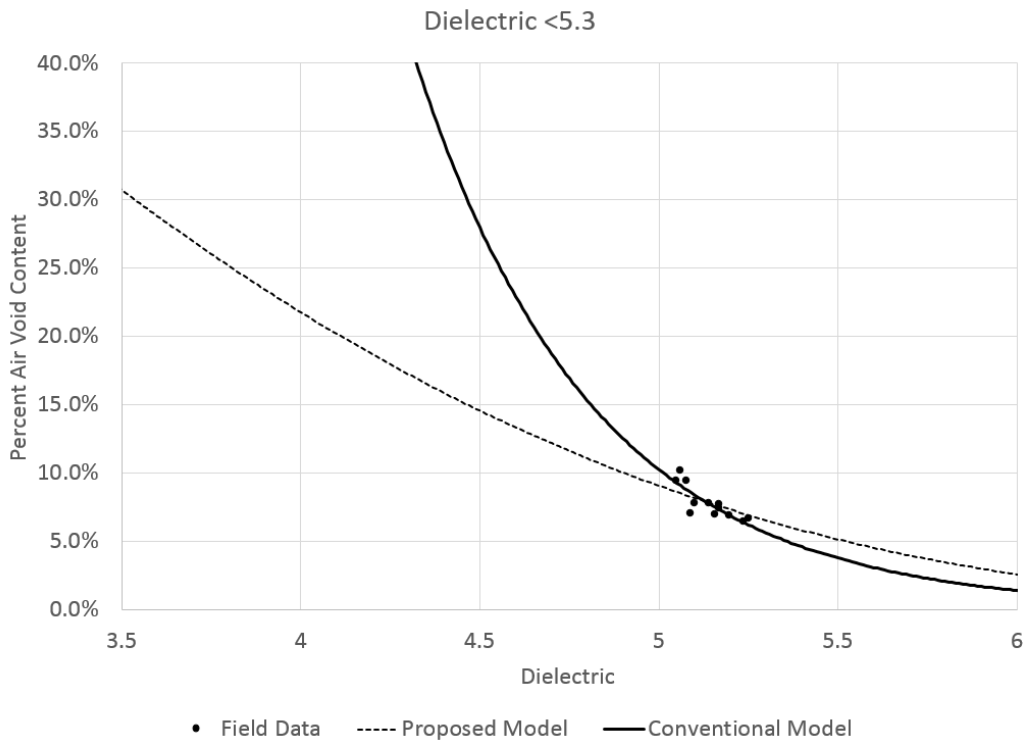


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2 Figure 4. Relationship between dielectric and air void content for example RDM projects (black)  
3 in comparison to the mix model equivalent from figure 2 (red).  
4



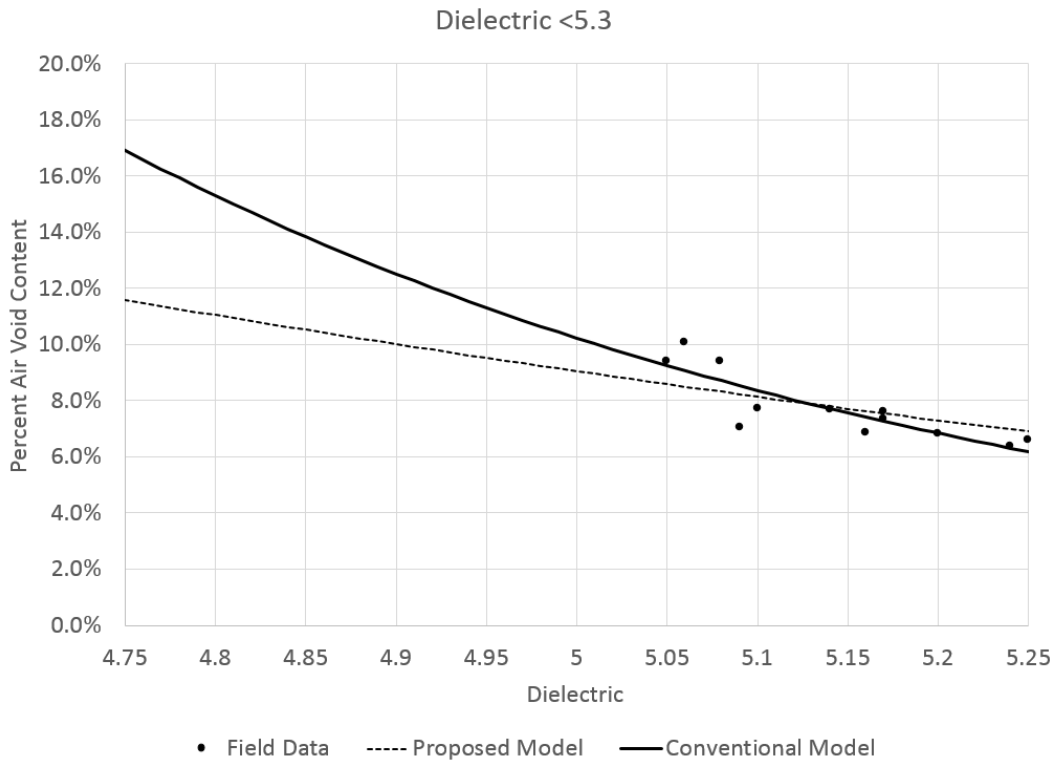
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Figure 5. Field collected data along with the currently accepted and proposed models.



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a. Full range of dielectrics



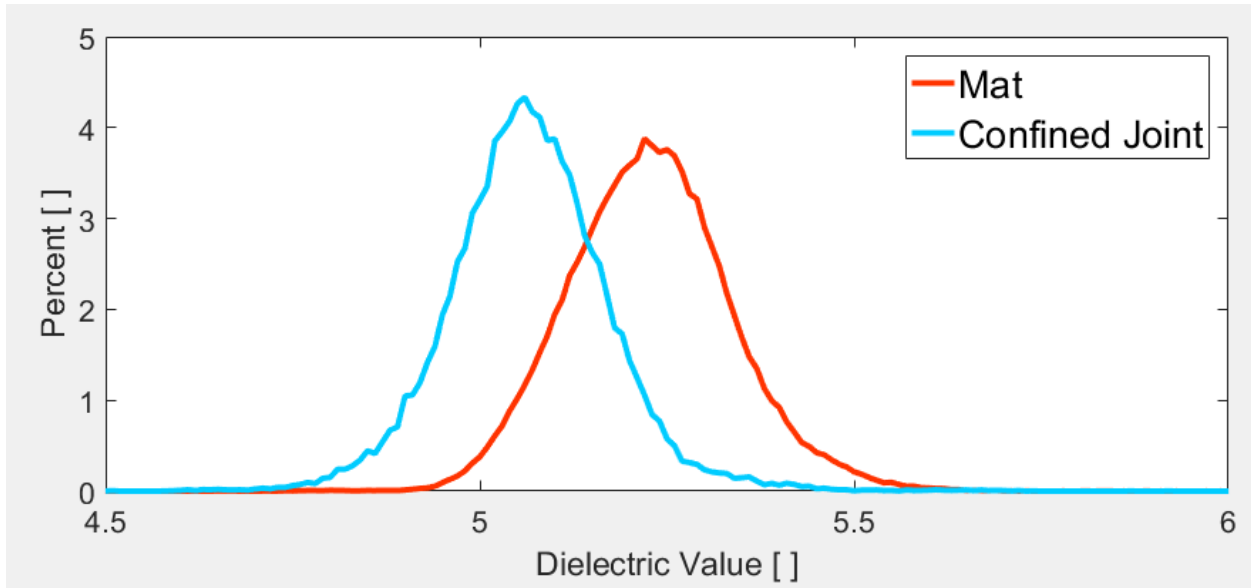
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b. Realistic range of dielectrics for confined joint

5 Figure 6. Comparisons of the predictions of the models calibrated with the field data and the  
6 modified dataset for a. conventional model and b. proposed model.

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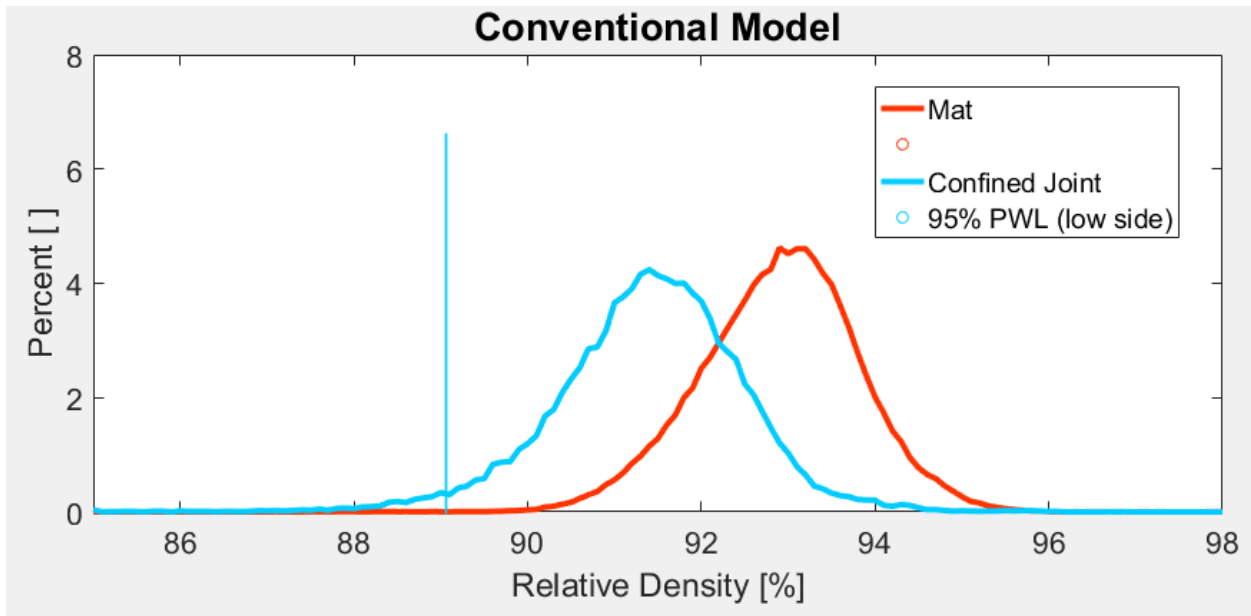
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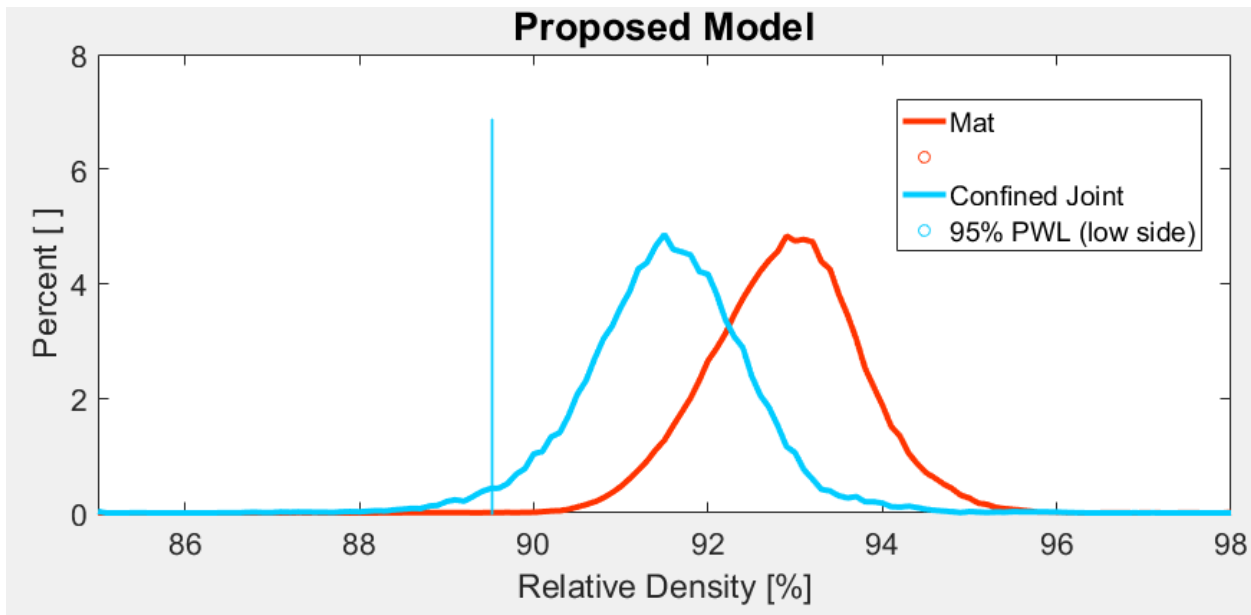
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Figure 7. Distribution of dielectric value measured over a 7 mile stretch of TH14 asphalt pavement.



1

2 a. Conventional model



3

4 b. Proposed model

5 Figure 8—Histograms of the relative densities for mat and confined joint of TH14, MN

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9 Figure 4. Relationship between dielectric and air void content for example RDM projects (black)  
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