Enhanced Model for Continuous Dielectric-Based Asphalt Compaction Evaluation

Kyle Hoegh, Corresponding Author
MnDOT Materials and Road Research
1400 Gervais Avenue, MS 645
Maplewood, MN 55109
Tel: 651-366-5526; Email: kyle.hoegh@state.mn.us

Shongtao Dai
MnDOT Materials and Road Research
1400 Gervais Avenue, MS 645
Maplewood, MN 55109
Tel: 651-366-5407; Email: shongtao.dai@state.mn.us

Trevor Steiner
MnDOT Materials and Road Research
1400 Gervais Avenue, MS 645
Maplewood, MN 55109
Tel: 651-366-5526; Email: trevor.steiner@state.mn.us

Lev Khazanovich
University of Pittsburgh
500 Pillsbury Drive S.E.
Minneapolis, MN 55455-0116
Phone: 612-624-4764; Email: lev.k@pitt.edu

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1
2 **ABSTRACT**
3 The compaction of asphalt concrete significantly affects long term pavement performance. While coring
4 provides a relatively accurate way of assessing in-place density at specific locations, the coverage of the
5 assessment is limited, especially at longitudinal joint locations. This can be particularly problematic since
6 it is difficult to identify problematic locations that are likely to fail prematurely using current compaction
7 assessment methods. Ground penetrating radars (GPR) provides an attractive non-destructive testing
8 alternative for evaluation of compaction quality, especially with recent significant improvements in the
9 GPR technology for this specific application. However, assessment of the air void content of the asphalt
10 mix from the GPR-measured dielectric constant of the surface requires conversion of dielectric variation to
11 air void content variation, which is the subject of this paper. An alternative to the commonly used model is
12 proposed, leading to more justifiable predictions for low values of dielectric constants. The proposed model
13 was used to interpret data from a 7-mile long asphalt overlay construction project. The results of the
14 interpretation as compared with the results obtained with the conventional model show an improvement on
15 the stability of the prediction at low air void contents, especially when core calibration data is limited and
16 uncertainty is considered. These results are promising in the direction of reducing field cores necessary to
17 have a stable model providing continuous compaction assessment of new asphalt pavement construction.
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25 *Keywords*: GPR, Compaction, Continuous, Assessment, NDT, Model
26

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_0, 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

$$e_r = \left(\frac{1 + \left(\frac{A_0}{A_i}\right)}{1 - \left(\frac{A_0}{A_i}\right)} \right)^2 \tag{1}$$

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

1
$$AV = A * exp(-B * \varepsilon)$$
Eq. 1

2 where, AV is the air void content, ε is the input hot mix asphalt dielectric constant, A and B are
3 calibration constants.
4
5 While this purely empirical relationship is effective in converting dielectric constant to air void
6 content when necessary dielectric precision is achieved and sufficient core calibrations are
7 available, a more rigorous approach should derive the air void- dielectric constant relationship for
8 the asphalt mix from the dielectric properties and volume fractions of the asphalt mix components.
9 While this ideal is not fully achievable using the currently available mix models, the mix-modeling
10 approach should be considered to ensure a more stable physics-based empirical model.
11
12 Electromagnetic mix-modeling has been developed for dielectric characterization of rock properties
13 such as estimation of rock porosity (*25*), water and clay influences (*26*), and other in geophysical
14 applications (*27*). This work was further extended for application to civil engineering materials
15 including evaluation of hot mix asphalt using ground penetrating radar (23, *28-30*). The mix model
16 method is based on the modelling of how an electromagnetic wave interacts with composite
17 materials using assumptions for how the wave interacts with the different asphalt concrete
18 components. The mix models evaluated in this study differ only in form, as the coefficients used
19 (dielectric and volumetric values) are the same in each model. A more detailed description of each
20 model can be found elsewhere (23, 28-30). Equations 2 through 4 show commonly accepted mix
21 models for asphalt concrete.
22
23

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

25

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

27

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

29
30
31 where, AV is the air void content, ε_{HMA} is the input hot mix asphalt dielectric constant, and the
32 remaining variables are given in Table 1.
33
34 The mixing models are advantageous as compared to a purely empirical approach in that they are
35 rational. These models – combined with laboratory testing and more accurate assumptions – may
36 eventually reduce or even eliminate the need for calibration cores. These models guarantee
37 reasonable trends, i.e. sensitivity to the properties of the individual components, so it may make
38 them less susceptible to errors. For example, each model can be verified for the extreme
39 conditions, such as predicting 100% air void content (0 bulk specific gravity) when a dielectric
40 constant of 1 is used. By comparison, the conventional purely empirical model does not hold true
41 outside of the measured dielectric ranges.
42
43 Al-Qadi et al. conducted a sensitivity study and evaluated the CRIM, Rayleigh, and Bottcher
44 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.
13
14
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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5
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 sensitivitiy (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 Transporatation 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 5th and 95th 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

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

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.
16
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3 **FIELD TESTING RESULTS**
4
5 As observed above, the conventional and modified model produce very similar results for a wide range of
6 the measured dielectric values, except when the dielectric values extend to low dielectric and air void
7 contents. While the conventional model leads to unreliable and physically inadmissible air void contents
8 for low dielectric values, the modified model offers a more rational and justifiable alternative.   To evaluate
9 the significance of this enhancement when applied to a full-scale field implementation, consider the data
10 collected with a RDM from 7 miles of a 1.5-in asphalt overlay of TH14 near Eyota, MN. Figure 7 shows
11 histograms of dielectric values measured with the RDM along a stretch of pavement and separated into 2
12 groups:

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

16
17 Using the conventional and modified calibration models, dielectric frequency distributions were
18 converted into the distributions of the relative densities, RD, defined as

19 $$RD = 1 - AV$$

20 Figure 8a shows the histograms obtained from these distributions with the conventional model
21 while Figure 8b shows the histograms obtained with the proposed model. Table 2 presents a
22 summary statistics obtained using these distributions. Comparisons of the Figures 8a and 8b
23 reveals that the relative density distributions for the mat are almost identical. Table 2 shows that
24 these distributions have the same mean and median values. The relative densities corresponding
25 to 95% percent within limits are also very similar: 91.0 and 91.2 percent from the conventional
26 and proposed models, respectively. The confined joint exhibited a lower relative density compared
27 to the mat compaction level, as expected. It is also not a surprise that a greater discrepancy between
28 the relative density distributions is observed for the confined joint. Although the mean and median
29 values are also very similar, the difference between the relative densities corresponding to 95%
30 percent within limits is significantly greater: 89.0 and 89.5 percent from the conventional and
31 proposed models, respectively. This suggests that the conventional model tends to overestimate
32 the percentage of the area with low compaction. Similar analysis of the unconfined side of the
33 joint showed an even greater difference between the models.
34
35
36 **CONCLUSIONS**
37
38 Early deterioration and long term performance of asphalt pavements is highly affected by quality
39 of compaction. To minimize potential delay in traffic closure on rehabilitation of heavily trafficked
40 areas where quality of compaction is especially important, it is often desirable that the
41 measurements can be taken over longer sections in a short timeframe immediately after final roller
42 compaction, while still providing the necessary pavement coverage. An accurate conversion of
43 the dielectric values to the air void content is especially important for the areas with low measured
44 dielectric values such as longitudinal joints, since they often cause early pavement failure.
45
46 This study shows the potential of compaction assessment using a SHRP2 recommended
47 technology referred to as the rolling density meter (RDM), which assesses compaction using a

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

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1 **Tables**

2

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

Model		Aggregate (εs)	Binder (εb)	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.

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

2

3

1 **Figures**
2
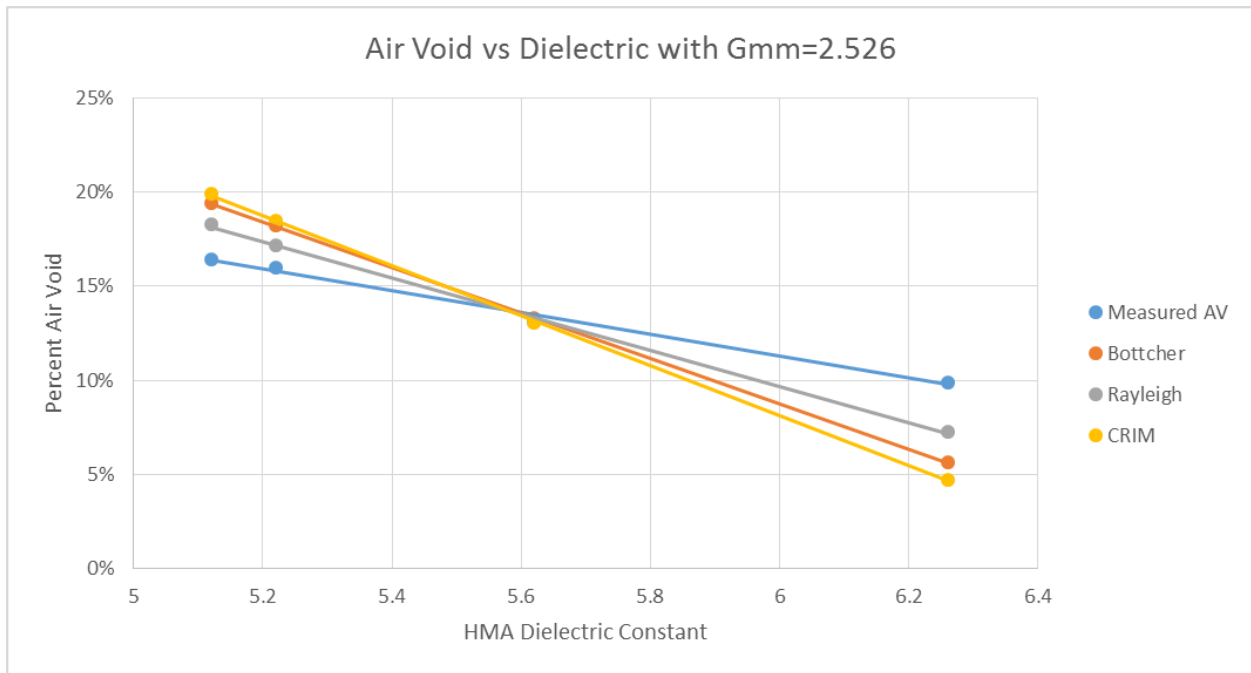
3
4 Figure 1. Measured (blue) versus predicted air void content using the CRIM, Rayleigh, and
5 Bottcher models (*23*).
6

1
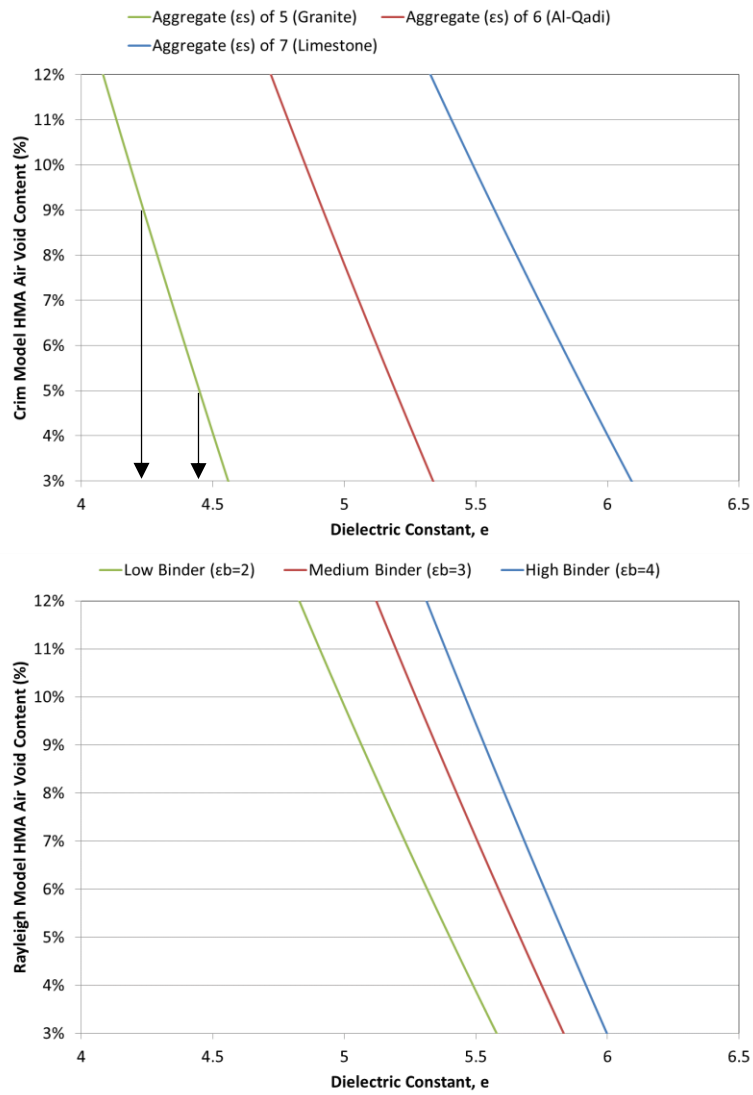
2
3   Figure 2. Predicted change in air void content using the Rayleigh model for a range in dielectric
4   values assumed for aggregate (top) and binder (bottom).
5

1
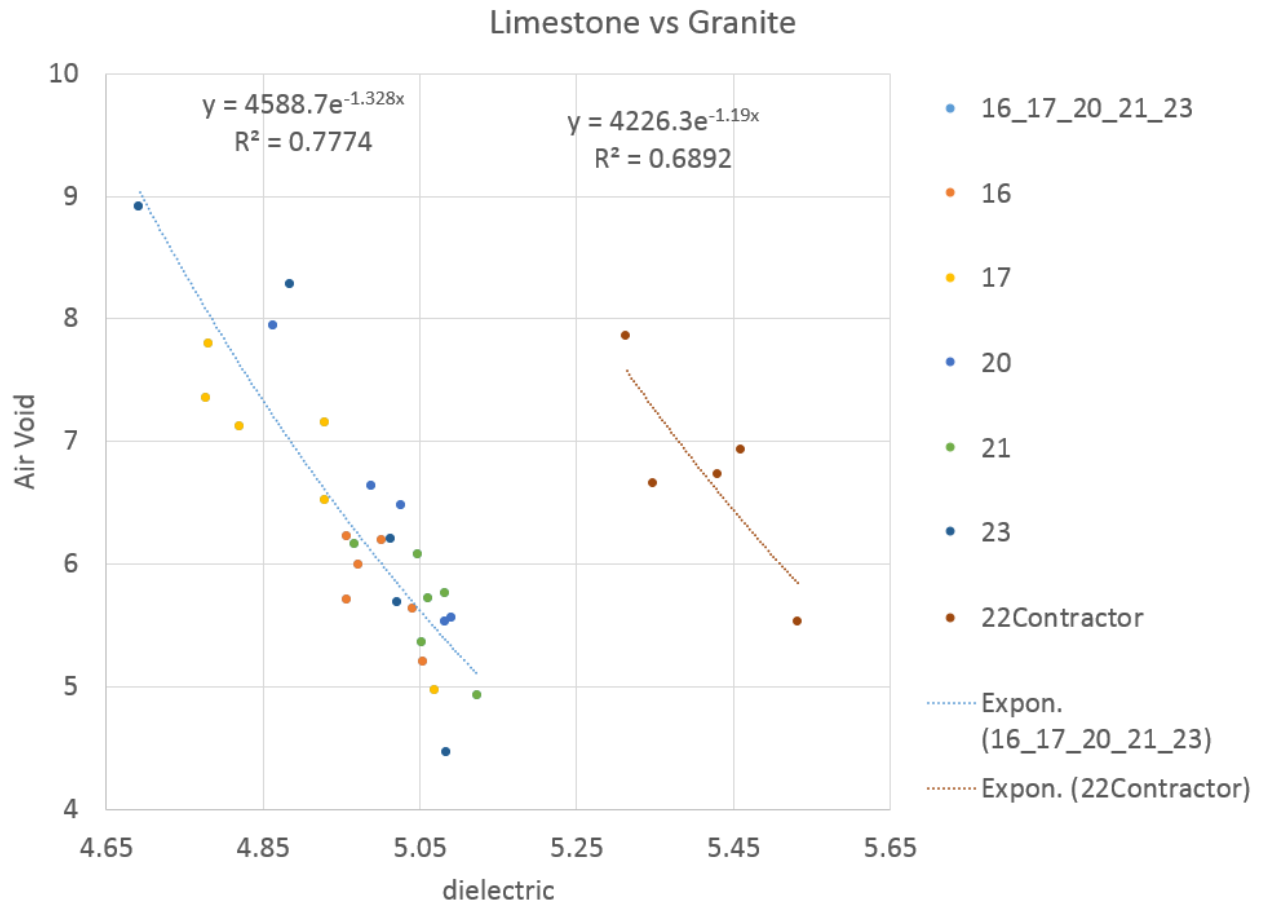
2
3
4   Figure 3 Example core versus dielectric data illustrating when different mixes can be combined
5   (left curve) and when they need to be separated (right curve).
6
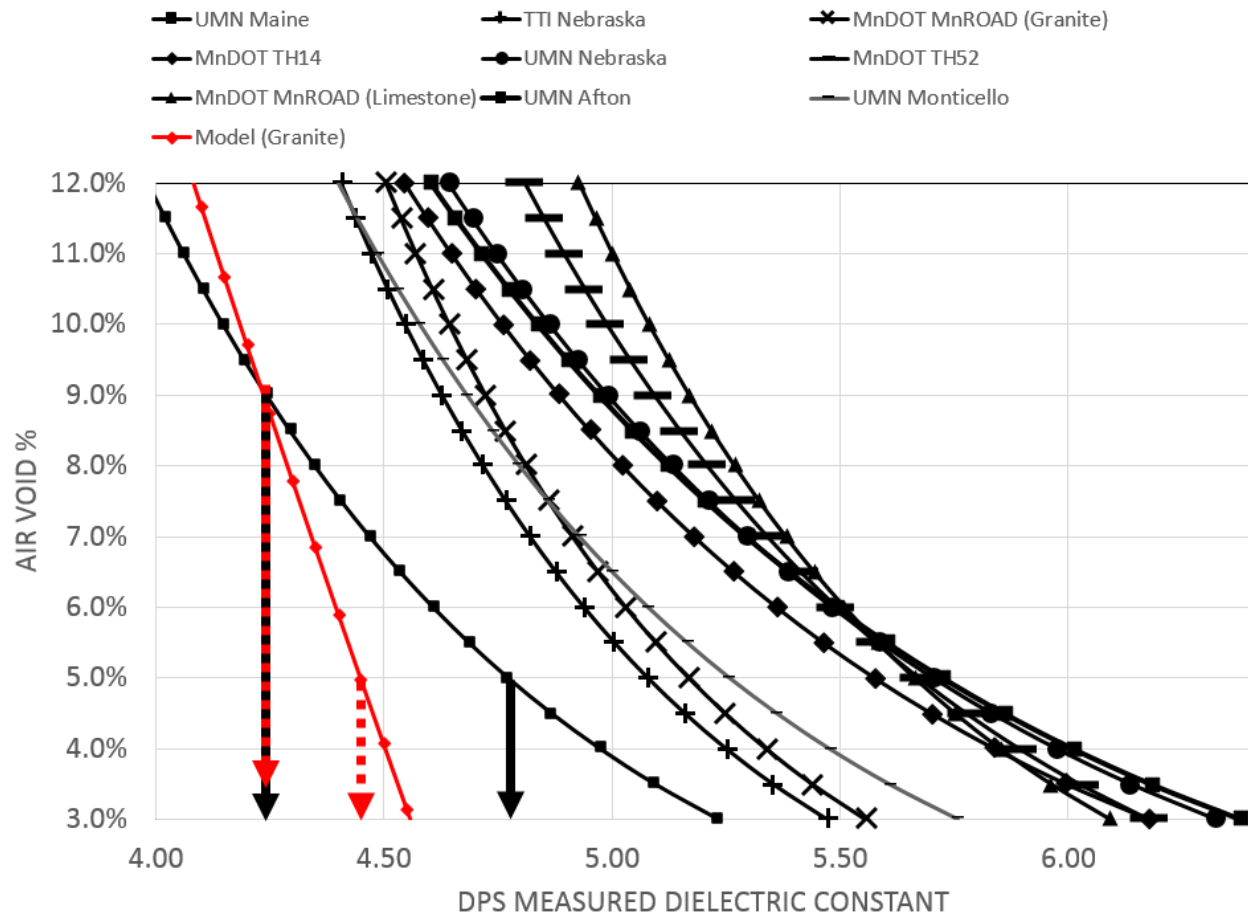
1
2   Figure 4. Relationship between dielectric and air void content for example RDM projects (black)
3   in comparison to the mix model equivalent from figure 2 (red).
4

All Data

$$AV = 8.491 * e^{-0.923*\delta}$$

$$AV = e^{-32.888\left(17.981\left|\frac{1}{e-15.770}-\frac{1}{1-15.770}\right|-1\right)}$$
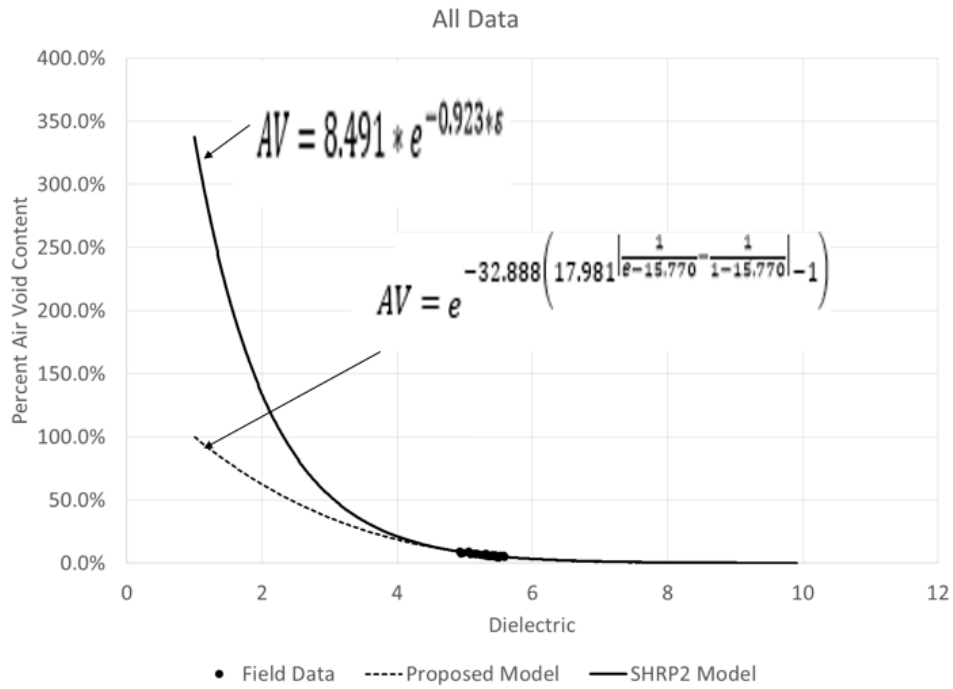
Figure 5. Field collected data along with the currently accepted and proposed models.
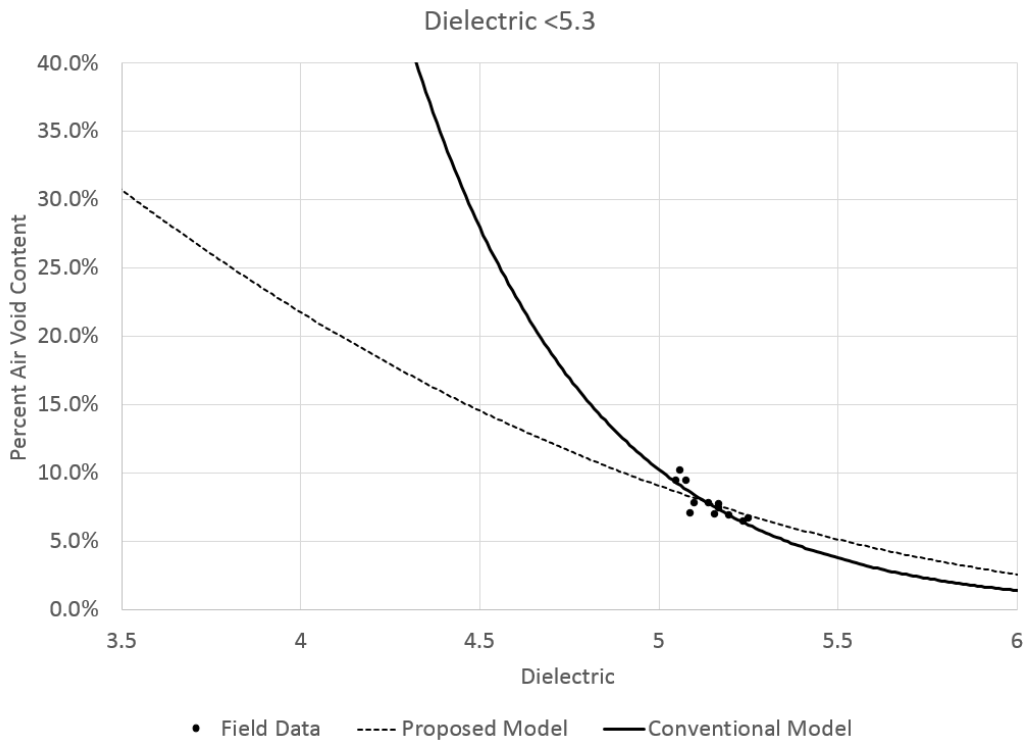
1
2   a.   Full range of dielectrics
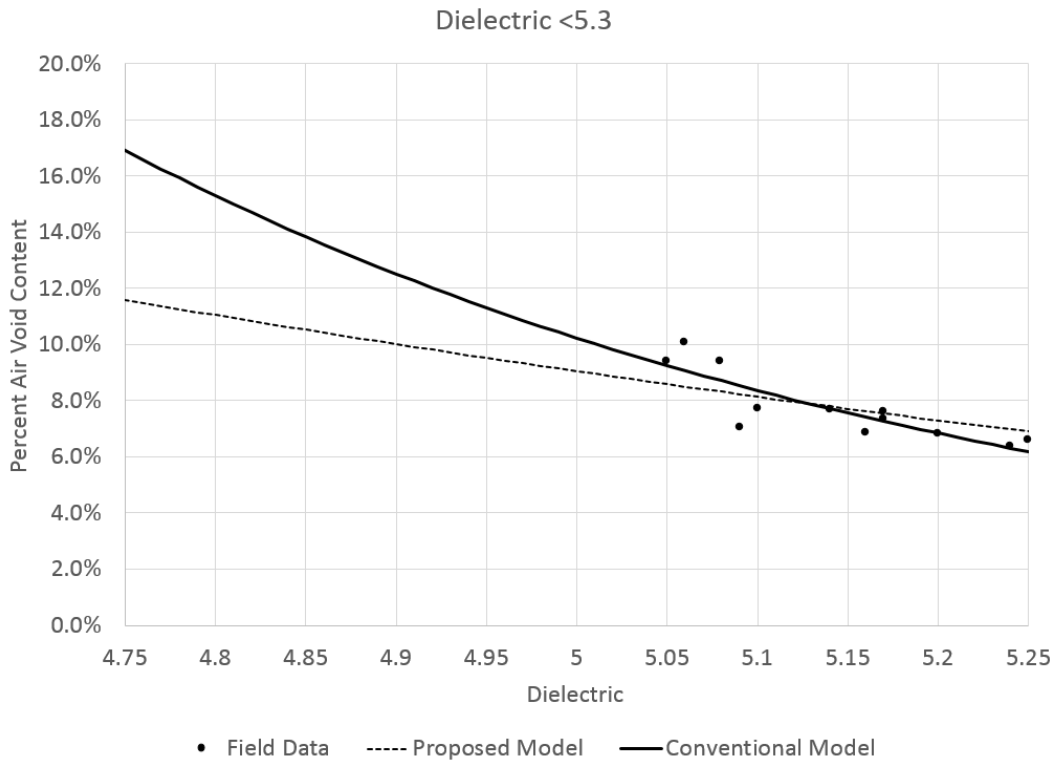
3
4   b.   Realistic range of dielectrics for confined joint

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.
7

1
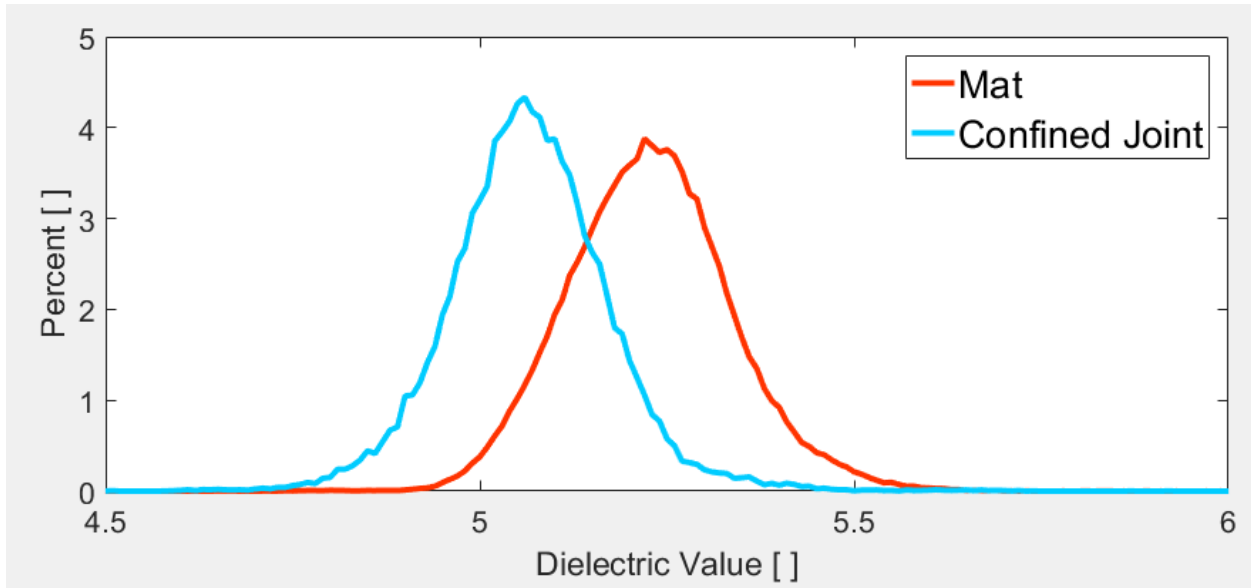
2
3   Figure 7.   Distribution of dielectric value measured over a 7 mile stretch of TH14 asphalt
4   pavement.
5
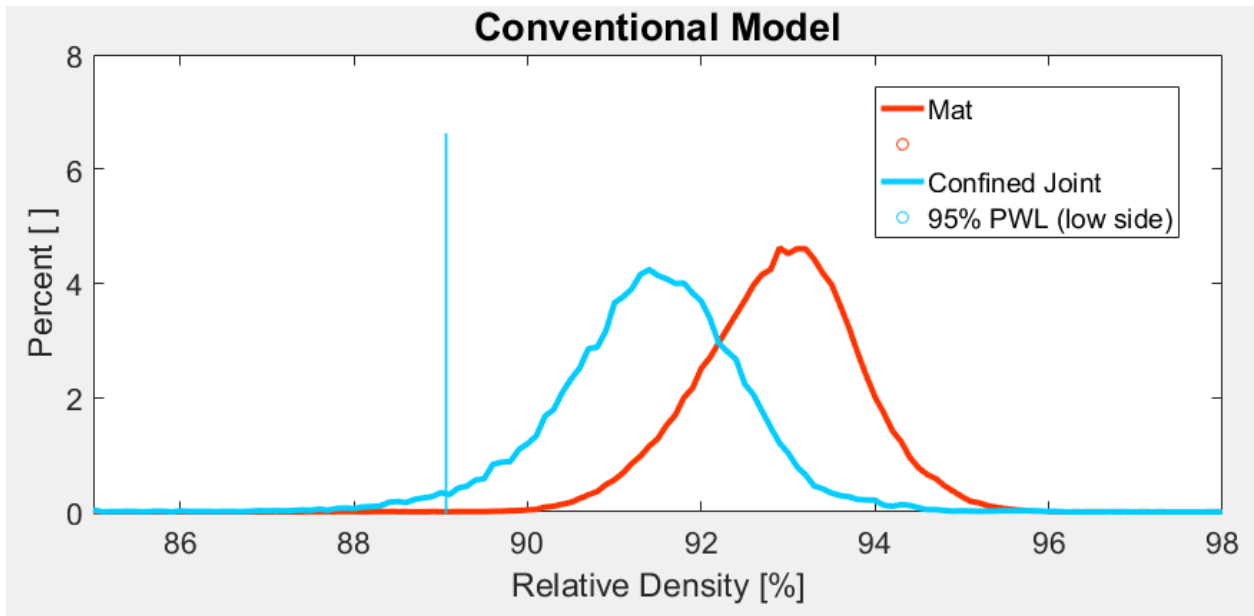
1

2   a.   Conventional model
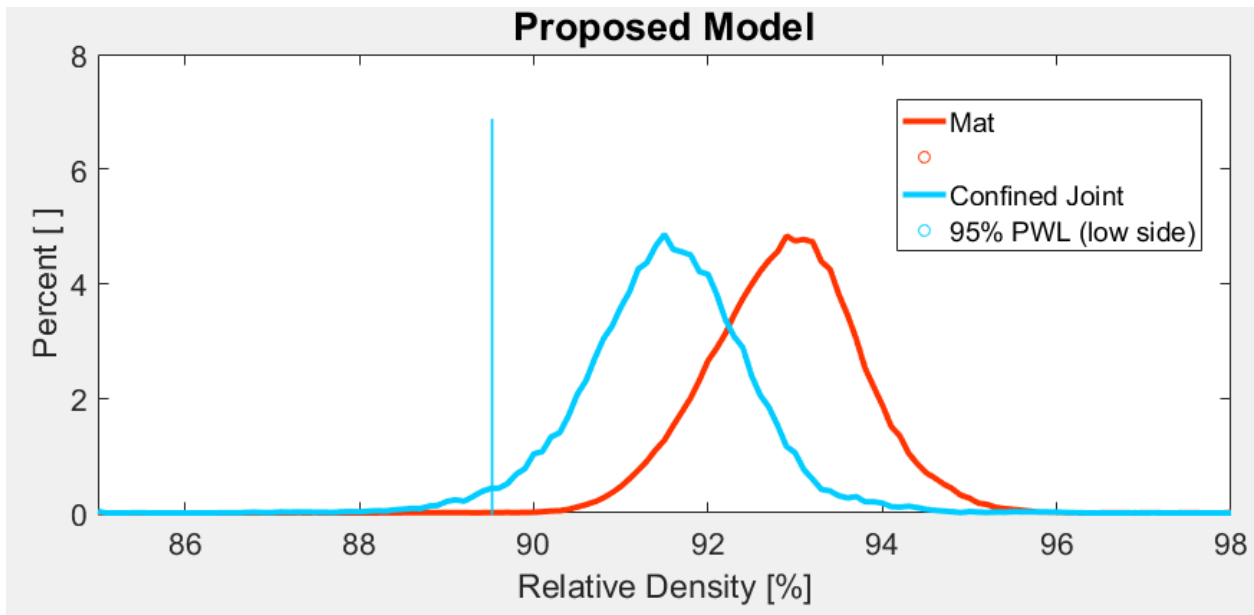
3

4   b.   Proposed model

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

6

1   **List of Tables**

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3   Table 1. Mix Model Inputs (23).

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

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